



Measuring the excellence contribution at the journal level: an alternative to Garfield's impact factor

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Abstract

The aim of this study is to analyze to which extent the Journal Impact Factor (JIF) reflects the amount of excellent publications contained in a journal in the corresponding subject category. We are introducing two percentile-based indicators in order to measure the excellence contribution at journal level. Calculations of these indicators have been carried out for twenty different Journal Citation Reports (JCR) subject categories to investigate the correlation with Garfield's Journal Impact Factor. Differences in the ranking according to all three indicators especially in First Quartile (Q1) of each category are shown and discussed. We have also studied the effect of multidisciplinary journals to the excellence contribution at category level and observed considerable differences between the twenty considered categories under analysis. Their omission would lead to neglect a large part of excellent publications, especially in the hard sciences. The introduced excellence indicators are very robust considering the types of documents considered for their calculation. The results of this study show that the journal excellence content is not completely reflected in the JIF measure and affects both Science and Social Science Edition categories. Therefore, the introduction of journal excellence indicators will help to give a more complete picture of the citation impact of a journal, because they are informing directly about the total and normalized excellence contribution of each journal to the corresponding subject category. They are a valuable enrichment and complement of the JIF, and more suitable than the Journal Citation Indicator, recently added to the JCR, because they reflect the inherent skewness of the citation distribution.

Keywords Journal impact factor · Journals · Excellence · Contribution · Indicators · Citation impact

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Introduction

Since its introduction by Garfield in the 1960s, first mentioned in 1963 (Garfield, 1972, 1976; Garfield & Sher, 1963), the Journal Impact Factor (JIF) is still one of the most common bibliometric indicators when it comes to measuring journal impact (Archambault & Larivière, 2009). Its popularity is unbroken, and not only because its introduction meant a revolution for the scientific community (Larivière & Sugimoto, 2019). The simple fact that, despite the development of a multitude of new indicators, none of the alternatives has prevailed testifies to the high acceptance of the JIF when it is used reasonably (Garfield, 2005; Gorraiz et al., 2020). The past has clearly shown that the JIF is not an all-in-one solution for various issues, which has led to controversial discussions and justified criticism (Alberts, 2013; Glänzel & Moed, 2002; Gorraiz, Gumpenberger, et al., 2012; Moed, 2002; Moed & van Leeuwen, 1995; Todorov & Glänzel, 1988). In response, several manifestos and statements were published especially due to the increasingly frequent misuse in research-assessment practices (American Society for Cell Biology [ASCB], 2012).

The first edition of the Journal Citation Reports (JCR)—including for the first time the Journal Impact Factor—was launched in 1976 and was based on the fundamental understanding that citations can be used as valuable criterion for the assessment of scientific journals (Garfield, 1976). The more frequently a journal is cited, the higher the recognition of its importance and prestige as information channel in its respective research field.

Researchers started to use the JIF in order to identify adequate publication venues and to optimize their publication strategies. As of its introduction editors and publishers rely on the JIF in order to estimate the reputation, prestige and market value of their journal portfolio. Furthermore, the JIF opened up a new support tool for librarians to back up decisions about subscriptions, to guarantee the presence of indispensable journals in their collections and to optimize their acquisition strategy. Finally, policy makers have thus gained a quantitative indicator for evaluation purposes, which additionally drove the expansion from its use for scientific information to application in evaluative contexts (cf. Glänzel, 2006).

The JIF has been further developed and improved over the years (extension of the citation window to 5 years, consideration of the journal self-citations, etc.) and nowadays a number of alternative journal citation-indicators are available such as the h-index for journals (Braun et al., 2006), eigenfactor metrics (Bergstrom et al., 2008; West et al., 2010), SJR (González-Pereira et al., 2010; Guerrero-Bote & Moya-Anegón, 2012), the SNIP indicator (Moed, 2010a, 2010b) or the CiteScore (cf. van Noorden, 2016). Nevertheless, the new edition of the JCR is eagerly expected each year, which shows the continuing importance of this analytical tool for the scholarly community and for research assessment.

Research assessment exercises are often performed for recent time periods. In these cases, impact analyses relying on citations are not very useful, because in many disciplines the citation window is practically too short for retrieving significant number of citations. Although it is not the appropriate indicator to measure the impact of a publication (Waltman & Traag, 2020), the JIF does provide a quick information on the impact and prestige of the journals in which the researcher, group or institution has been able to publish. Being published in journals with high JIF is much more difficult (higher rejection quotes), and successful publication in these journals provides recognition. JIF

also helps to identify the top journals in each field according to their impact or prestige. This is why the JIF plays such a key role. The competition to be included in the Web of Science (WoS) Core Collection and to be indexed as a journal or to publish in one continues unabated (Osterloh & Frey, 2014) and is inextricably linked to the question of how the citation impact and prestige of a journal is measured.

However, since the introduction of the JIF, many analytical tools have been developed and are available, enabling a very quick and automatic calculation of the percentiles of the most cited publications for each publication year and each subject category (Lozano et al., 2012). Nowadays the normalized citation counts like Category Normalized Citation Impact (CNCI) and the number and percentage of Top 10% and Top 1% most cited publications are essential indicators in citation analyses (Adams et al., 2007; Gorraiz & Gumpenberger, 2015; Gorraiz, Reimann, et al., 2012). In addition, in June 2021 Clarivate analytics presents the 2021 JCR offering a revamped user interface with new interactive graphics that permit a more complete, dynamic view of data. The new JCR also included a new indicator, the Journal Citation Indicator (JCI),¹ that it is the journal-level CNCI and has been designed to complement the JIF.

Therefore, it can be quite interesting to use these normalized indicators as an alternative to the JIF. Does the JIF reflect the amount of excellent publications contained in a journal or in a subject category? Are there other approaches to paint a more precise picture of journal excellence? This is the subject of our study. A previous study focusing on five selected WoS subject categories was presented at the 18th International Conference on Scientometrics and Informetrics (ISSI 2021) and published in the corresponding proceedings (Gorraiz et al., 2021). This study has now been expanded to a total of 20 WoS subject categories for this new paper.

Research questions

The main objective of this article is to present new indicators to measure scientific excellence that can complement the JIF designed to provide a robust and size-independent journal performance measure. To achieve this objective, we have established the following scientific questions, which we have classified into two blocks. In the first, we include questions related to the design of the indicators of excellence and their relationship with other bibliometric indicators. These research questions are the following:

1. Is it possible to design complementary and alternative indicators for scientific journals considering the number of highly cited publications in the JCR categories?
2. Could these new indicators of research excellence based on a percentile approach supplement the JIF as an improved assessment of the citation impact of a journal?
3. How does the JIF and other bibliometric indicators correlate with the proposed percentile-based indicators?

Once the indicators have been designed and presented, two research questions have been defined. The aim of these two questions is to find out the inner workings of the indicators.

¹ <https://clarivate.com/blog/introducing-the-journal-citation-indicator-a-new-field-normalized-measurement-of-journal-citation-impact/>.

4. An issue when calculating journal indicators is not to consider categories as closed lists. In the JCR, the category lists only include journals that publish in that category, however, when counting citations, the citations received from all journals are included. Considering this, how do multidisciplinary journals (e.g., journals of the WoS category “Multidisciplinary Science”), affect the indicators and JCR calculations? We refer to the effect of papers published in multidisciplinary journals (Plos One, Nature, Science, ...) but belonging to a specific field according to Incites recategorization.
5. There is an asymmetry in the calculation of the JIF. In the numerator, the citations to all types of documents are summed up, while in the denominator only research articles and reviews are considered. Considering this, how sensitive are our indicators to the choice of document types, particularly of the so-called ‘citable items’ (i.e., articles and reviews) instead of all document types?

The article is organized as follows: (1) in the methodology we will give specific details of the calculation of the indicators. In the results (2) we provide a descriptive overview of the 20 categories analyzed, and (3) a more detailed case study of the indicators of excellence applied exclusively to the category of “Library and Information Science” (ILS) will be presented. (4) In the next section, a summary of the correlations of the indicators of excellence in the 20 categories is presented. Finally, (5) in separate sections the results concerning the effect of document typologies and multidisciplinary journals are analyzed.

In order to provide more information and ensuring the reproducibility and validity of the data this paper is complemented by the following materials deposited in the Zenodo repository: (1) The work in progress presented at the ISSI 2021 conference (<https://doi.org/10.5281/zenodo.5679387>). (2) The dataset with all the collected data distributed in five tab-separated values (TSV) files (<https://doi.org/10.5281/zenodo.5676184>). Finally, the video of Juan Gorraiz’s oral presentation at the ISSI is available at https://www.youtube.com/watch?v=Imwryb_pNhk.

Methodology

All documents assigned to the 20 selected WoS subject categories published of the years between 2009 and 2018 were retrieved in InCites, excluding ESCI documents since to identify the journals we used the sources of the publications that had a JIF associated with them. Table 1 lists the 20 WoS subject categories considered in this study according to JCR. The categories were chosen to give a view as broad as possible of the various publishing cultures.

In this study, we are considering only journals with a JIF, and we are performing the analyses for two different groups:

- Group 1: only journals assigned to each WoS subject category according to JCR (“JCR Cat.”).
- Group 2: including all multidisciplinary journals that, according to InCites, have likewise contributed to this category (“JCR Cat. + Multidisciplinary”).²

² The category “Multidisciplinary Sciences” does not exist in InCites and the publications attributed to it in the WoS Core Collection are redistributed according to their topic in the more specific categories.

Table 1 List of selected JCR categories, their abbreviations and JCR edition

Abbreviation	Category name in JCR	JCR edition
BUS	Business	SSCI
CHEA	Chemistry, Analytical	SCIE
CHEM	Engineering, Chemical	SCIE
CHPH	Chemistry, Physical	SCIE
COMP	Computer Science, Artificial Intelligence	SCIE
ECO	Economics	SSCI
ECOL	Ecology	SCIE
EDU	Education & Educational Research	SSCI
ENV	Environmental Sciences	SCIE
HIS	History	SSCI
ILS	Information Science & Library Science	SSCI
MICR	Microbiology	SCIE
NEUR	Neurosciences	SCIE
NURS	Nursing	SSCI
PHAR	Pharmacology & Pharmacy	SCIE
PHCM	Physics, Condensed Matter	SCIE
POLS	Political Science	SSCI
PSYC	Psychology	SCIE
STPR	Statistics & Probability	SCIE
VIR	Virology	SCIE

For each journal, we list:

- Number of publications published in this journal in JCR Cat.: $p(J)$.
- Number of excellent publications published in this journal in JCR Cat.: $x(J)$.
 For each category we list:
 - Total number of publications in JCR Cat.: $p(T)$.
 - Total number of excellent publications in JCR Cat.: $x(T)$.

In this study the term “excellent publications” or “excellence” is used as synonym for publications belonging to the Top 10% most cited documents in the same JCR category, publication year and document type.

Beside the JIF retrieved from the JCR Edition 2020, we have calculated the following indicators for each journal:

1. Journal Percentage of Excellent Publications (JPEP) = $(x(J)/p(J))$ = Number of excellent publications published in this journal in the PY = 2009–2018 in this WoS category/Total number of publications published in this journal in the PY = 2009–2018 in this WoS Category.
2. Journal Contribution to the Excellence of the Category (JCEC) = $(x(J)/x(T))$ = Number of excellent publications published in this journal in the PY = 2009–2018 in this WoS category/Total number of excellent publications published in the PY = 2009–2018 in this WoS category.

Both indicators are size dependent: The first one (JPEP) can reach very high values for journals with just few publications in the category, and the second one (JCEC) benefits journals from a large number of publications. Therefore, we have also calculated two further indicators:

3. Journal Brute Excellence (JBE) = $100 \times \text{JPEP} \times \text{JCEC} = 100 \times x^2(J)/(p(J) \times x(T))$.
4. Journal Normalized Excellence (JNE) = $(x(J)/x(T))/(p(J)/p(T))$ = Journal Contribution to the Excellence (JCEC)/Journal Contribution to the Category.

The first one reflects the total brute excellence force or brute contribution of the journal to the category. The second one provides the normalized excellence contribution of the journal to the category. Together they provide a more complete picture of the journal excellence. We are using the JNE especially for the analysis limited to the journals assigned to the JCR category under study (“JCR Cat.”), because the number of publications of these journals is significant, resulting in relevant JNE values. Note that JNE is inspired by the “Attractivity Index” by Schubert and Braun (1996), which is, in turn, defined based on the model of the Activity Index introduced into scientometrics by Frame (1977). Both indicators have been used since the late 1980s to reflect a country’s, region’s or other unit’s relative contribution to research productivity and citation impact in given subject fields (cf. Schubert et al., 1989). JNE here expresses a journal’s contribution to the excellence in a given subject. As such JNE, analogously to the above-mentioned indicators by the Hungarian research group, is a balance measure with neutral value 1, i.e., a journal contributes relatively more (less) to the subject’s excellence according as $\text{JNE} > (<) 1$. It is not contributing at all, if $\text{JNE} = 0$. The only conceptual deviation of JNE from activity/attractivity is that the balance is not considered across subjects but across units (i.e., journals). A consequence of the “balance” property of this concept is that not all journals can contribute relatively more (less) than expected—some journals assigned to the subject category reflect relatively more excellence than the subject standards, others contribute to subject excellence to a lesser extent.

When analyzing the effect of the multidisciplinary journals, we use the JBE. High impact multidisciplinary journals (like *Nature* or *Science*) contributing with rather few publications to the category could yield high JNE values, but according to the JBE no significant contributions are achieved. Pearson Correlations were then performed for the JIF, JBE and JNE for the 20 categories considering only journals assigned to the subject category (“JCR Cat.”) and including also multidisciplinary journals (“JCR Cat. + Multidisciplinary”). Furthermore, we have compared the Q1 journals assigned to each category according to JCR Edition 2020 with the Top Journals according to the two new indicators JNE (“JCR Cat.”), and JBE (“JCR Cat. + Multidisciplinary”).

In order to address research question 4, we have analyzed and discussed the contribution of other journals not directly assigned to the corresponding category, like e.g., the multidisciplinary journals, to the excellence of the category. For this purpose, we have introduced two more indicators:

5. Category Percentage of Multidisciplinarity (CPM) = Number of publications added by multidisciplinary journals not directly assigned to this category according to the JCR (e.g., *Nature*, *Science*, *PLOS ONE*, etc.)/Total number of publications in the category.
6. Category Excellence Degree Multidisciplinarity (CEDM) = Number of excellent publications added by journals not directly assigned to this category according to the JCR (e.g., *Nature*, *Science*, *PLOS ONE*, etc.)/Total number of excellent publications in the category.

Finally, we have also performed our analysis not only for the document types articles and reviews, but also for all document types in order to address research question 5.

Results

General overview

Table 2 gives an overview of the number of journals, publications and excellent publications for each category considered in this study. The differences between all document types (All types) and article and reviews (Art./Rev.) are also considered. The table makes clear that the selected categories represent very different communities and research categories: “Physics, Condensed Matter” (PHCM), “Virology” (VIR) or “History” (HIS) stand for small document sets and small scientific communities. On the other hand, we have large categories with a considerable number of journals such as “Economics” (ECO), “Neurosciences” (NEU) or “Pharmacology & Pharmacy” (PHAR).

With the aim of providing an overview of the categories and their characteristics Table 3 shows the Category Percentage of Multidisciplinarity (CPM) and the Category Excellence Degree Multidisciplinarity (CEDM), as well as the contribution of other document types than articles or reviews to the total excellence in the category on a percentage basis. The category percentage and degree of multidisciplinarity are very different according to the subject categories. The highest values are reported by the categories “Ecology” (ECOL), “Statistics & Probability” (STPR) and “Microbiology” (MICR), followed by “Virology” (VIR), “Neurology” (NEUR) and “History” (HIS). More than 13% of the excellent publications are published in multidisciplinary journals in the category of “Ecology” (ECOL) and around 10% in the category “History”. However, in “Education (EDU), “Pharmacology and Pharmacy” (PHCM), “Political Sciences” (POLS), “Business” (BUS), “Economics” (ECO) and “Chemistry, Analytical” (CHEM) the effect of the multidisciplinary journals is almost inexistent or very low. In “Information Science & Library Science” (ILS) the effect is much higher in the total number of publications (CPM) than in the number of excellent publications (CEDM) as well as for articles and reviews as in comparison to all document types. These results show that the effect of multidisciplinary journals can affect both science and social science categories but with different intensity.

Articles and reviews are mostly responsible for the number of excellent publications in all categories. This is even true for the categories related to the Social Sciences where big differences between the total number of all document types compared to articles and reviews can be observed (see Table 2). However, the results compiled in Table 4 show that the consideration of other document types than articles and reviews may be of significance in some categories of the Science (SCIE) as well as of the Social Sciences Edition (SSCI). Articles and reviews are mostly responsible for the number of excellent publications in all categories. This is even true for the categories related to the Social Sciences where big differences between the total number of all document types compared to articles and reviews can be observed (see Table 2). The lowest percentage of articles and reviews within the excellent publications is observed for “Neurosciences” (NEUR), “Psychology” (PSYC) and “Nursing” (NURS) with around 73%. Other document types than articles and reviews (especially editorial materials and letters) are responsible for almost a fourth of the excellence in these three categories. These categories are followed by “Pharmacology and Pharmacy” (PHCM), “Information Science & Library Science” (ILS) and “History” (HIS) with

Table 2 Overview of the 20 categories analyzed in this study including number of publications considering JCR categories and multidisciplinary journals as well as articles and reviews versus other document typologies

Categories	Document types	JCR cat			JCR cat. + multidisciplinary		
		Nr. journals	Nr. pubs	Nr. exc. pubs	Nr. journals	Nr. pubs	Nr. exc. pubs
BUS	All types	148	79.199	13.527	182	80.653	13.541
	ART./REV	142	64.757	12.200	176	67.482	12.209
CHEA	All types	81	238.126	24.242	150	242.854	24.447
	ART./REV	78	221.925	23.229	145	228.313	23.424
CHEM	All types	123	314.726	31.800	183	325.623	31.850
	ART./REV	119	278.877	30.404	180	293.356	30.445
CHPH	All types	143	568.243	57.500	208	588.733	58.642
	ART./REV	141	553.070	56.325	206	575.062	57.386
COMP	All types	131	119.885	14.347	188	122.688	14.512
	ART./REV	131	115.437	13.959	184	118.122	14.105
ECO	All types	351	242.035	29.987	433	246.764	30.481
	ART./REV	343	181.852	26.332	424	187.937	26.694
ECOL	All types	146	175.811	18.327	222	204.297	21.168
	ART./REV	144	164.122	17.380	220	190.644	19.959
EDU	All types	258	125.063	24.607	319	126.217	24.695
	ART./REV	256	101.428	21.862	305	102.839	21.917
ENV	All types	245	437.126	45.977	371	454.868	47.157
	ART./REV	245	412.574	43.870	364	428.860	44.853
HIS	ALL TYPES	99	82.575	8.983	126	88.258	9.902
	ART./REV	99	25.443	7.371	119	27.992	8.157
ILS	ALL TYPES	79	92.657	7.077	144	95.329	7.213
	ART./REV	73	33.745	5.850	131	37.828	5.950
MICR	All types	123	219.115	22.090	293	240.188	25.347
	ART./REV	122	190.814	19.841	290	210.817	22.942
NEUR	All types	255	562.803	50.383	429	601.928	56.192
	ART./REV	252	357.659	35.444	425	395.276	40.876
NURS	All types	120	115.069	11.349	210	1189.38	11.991
	ART./REV	118	71.480	8.292	201	75.018	8.811
PHART./REV.	All types	263	497.369	49.323	436	515.079	50.445
	ART./REV	260	357.481	39.256	435	37.3846	40.193
PHCM	All types	63	285.812	29.762	100	289.678	29.934
	ART./REV	62	278.819	29.145	98	283.149	29.307
POLLS	All types	177	111.552	16.483	204	11.3160	16.624
	ART./REV	170	60.863	14.116	198	65.477	14.233
PSYC	All types	70	106.215	9.672	110	108.410	9.719
	ART./REV	69	63.790	7.046	103	65.916	7.092
STPR	All types	123	87.033	7.910	192	95.658	10.601
	ART./REV	122	82.054	7.540	185	90.577	10.172
VIR	All types	35	83.006	7.269	145	89.138	8.248
	ART./REV	34	65.277	6.258	140	71.280	7.174

Table 3 Category Percentage of Multidisciplinarity (CPM) and the Category Excellence Degree Multidisciplinarity (CEDM) for the 20 subject categories according to InCites

Edition	Categories	All documents types		Articles and reviews	
		Category percentage of multidis- ciplinarility (CPM) (%)	Category excellence degree multi- disciplinarity (CEDM)	Category percentage of multidis- ciplinarility (CPM) (%)	Category excellence degree multidisciplinarity (CEDM)
SCIE	ECOL	13.94	13.421	13.91	12.921
SCIE	STPR	9.02	25.384	9.41	25.875
SCIE	MICR	8.77	12.85	9.49	13.517
SCIE	VIR	6.88	11.87	8.42	12.768
SCIE	NEUR	6.50	10.338	9.52	13.289
SSCI	HIS	6.44	9.281	9.11	9.636
SCIE	ENV	3.90	2.502	3.80	2.192
SCIE	CHPH	3.48	1.947	3.82	1.849
SCIE	PHAR	3.44	2.224	4.38	2.331
SCIE	CHEM	3.35	0.157	4.94	0.135
SSCI	NURS	3.25	5.354	4.72	5.89
SSCI	ILS	2.80	1.885	10.79	1.681
SCIE	COMP	2.28	1.137	2.27	1.035
SCIE	PSYC	2.02	0.484	3.23	0.649
SCIE	CHEA	1.95	0.839	2.80	0.832
SSCI	ECO	1.92	1.621	3.24	1.356
SSCI	BUS	1.80	0.103	4.04	0.074
SSCI	POLS	1.42	0.848	7.05	0.822
SCIE	PHCM	1.33	0.575	1.53	0.553
SSCI	EDU	0.91	0.356	1.37	0.251

Table 4 Percentage of excellence contribution from other document types than articles and review considering the presence or absence of multidisciplinary journals

Edition	Categories	JCR cat	JCR cat. + multidisciplinary
		% excellence contribution from other document types than “Art. + Rev.”	% excellence contribution from other document types than “Art. + Rev.”
SCIE	NEUR	29.65	27.26
SCIE	PSYC	27.15	27.03
SSCI	NURS	26.94	26.52
SCIE	PHAR	20.41	20.32
SSCI	HIS	17.95	17.62
SSCI	ILS	17.34	17.51
SSCI	POLS	14.36	14.38
SCIE	VIR	13.91	13.02
SSCI	ECO	12.19	12.42
SSCI	EDU	11.16	11.25
SCIE	MICR	10.18	9.49
SSCI	BUS	9.81	9.84
SCIE	ECOL	5.17	5.71
SCIE	STPR	4.68	4.05
SCIE	ENV	4.58	4.89
SCIE	CHEM	4.39	4.41
SCIE	CHEA	4.18	4.18
SCIE	COMP	2.70	2.80
SCIE	PHCM	2.07	2.09
SCIE	CHPH	2.04	2.14

Other are non “articles & reviews” to the category excellence for (1) only journals assigned to each WoS subject category according to Journal Citation Reports (“JCR Cat.”), and (2) including all multidisciplinary journals that, according to InCites, have likewise contributed to this category (“JCR Cat. + Multidisciplinary”)

around 80%, followed by “Virology” (VIR) and “Economics” (ECO) with around 88%. The highest percentage of articles and reviews within the excellent publications is reported in in “Chemistry, Physical” (CHPH), “Physics, Condensed Matter” (PCM) and “Computer Science, Artificial intelligence” (COMP) with almost 98%. In this study, we are focusing on the document types: articles (Art.) and reviews (Rev.). In “[Final remarks](#)” section”, the effect of the document types will be further analyzed and discussed.

Case study: “Information Science & Library Science” (ILS)

In order to offer a first approximation of the indicators, a first study has been carried out by applying them to the category of “Information and Library Science” (ILS), Table 5 provides an example of the results obtained for this and includes all the indicators mentioned in the methodology. The table shows only the First Quartile (Q1) journals according to the JIF. Figure 1 shows the correlation between the JIF and the two new indicators for all journals of the category “Information Science & Library Science” (ILS). The correlation is rather moderate (JBE; $r=0.763$, see Table 3), most notably for the normalized JNE ($r=0.906$, see Table 3), but some of the journals change their position, if a normalized

Table 5 Excerpt of the table of indicators calculated in “Information Science & Library Science” (ILS), only Q1 journals according to the JIF 2019 (article and reviews, published between 2009 and 2018)

	Journal impact factor	Nr pubs	Nr. excellent pubs	% Journal contrib IJEP	% Category contrib JCEC	Final indicators	
						JBE Brute excellence	JNE norma. excellence
Int. Journal of information management	8.210	796	357	44.85	6.10	2.737	2.851
MIS Quarterly	5.370	511	333	65.17	5.69	3.709	4.143
Journal of Computer-Mediated Communication	5.366	338	154	45.56	2.63	1.199	2.897
Journal of Strategic Information Systems	5.231	192	79	41.15	1.35	0.556	2.616
Information & Management	5.155	640	294	45.94	5.03	2.309	2.921
Government Information Quarterly	5.098	598	231	38.63	3.95	1.525	2.456
Information Processing & Management	4.787	685	151	22.04	2.58	0.569	1.401
Journal of Knowledge Management	4.745	663	264	39.82	4.51	1.797	2.532
Journal of Informetrics	4.611	734	204	27.79	3.49	0.969	1.767
Information systems journal	4.188	247	83	33.60	1.42	0.477	2.136
Telematics and Informatics	4.139	702	195	27.78	3.33	0.926	1.766
Journal of the American Medical Informatics Association	4.112	1713	525	30.65	8.97	2.750	1.948
MIS Quarterly Executive	4.088	163	32	19.63	0.55	0.107	1.248
Int. J. Computer-supported collaborative learning	4.028	194	55	28.35	0.94	0.267	1.802
Journal of Management Information Systems	3.949	406	157	38.67	2.68	1.038	2.458
Int. Journal of Geographical Information Science	3.733	1079	222	20.57	3.79	0.781	1.308
Journal of Information Technology	3.625	217	56	25.81	0.96	0.247	1.641
Information Systems Research	3.585	497	204	41.05	3.49	1.431	2.610
Information and Organization	3.300	132	29	21.97	0.50	0.109	1.397
Journal of the Association for Information Systems	2.957	317	89	28.08	1.52	0.427	1.785
Scientometrics	2.867	2921	495	16.95	8.46	1.434	1.077

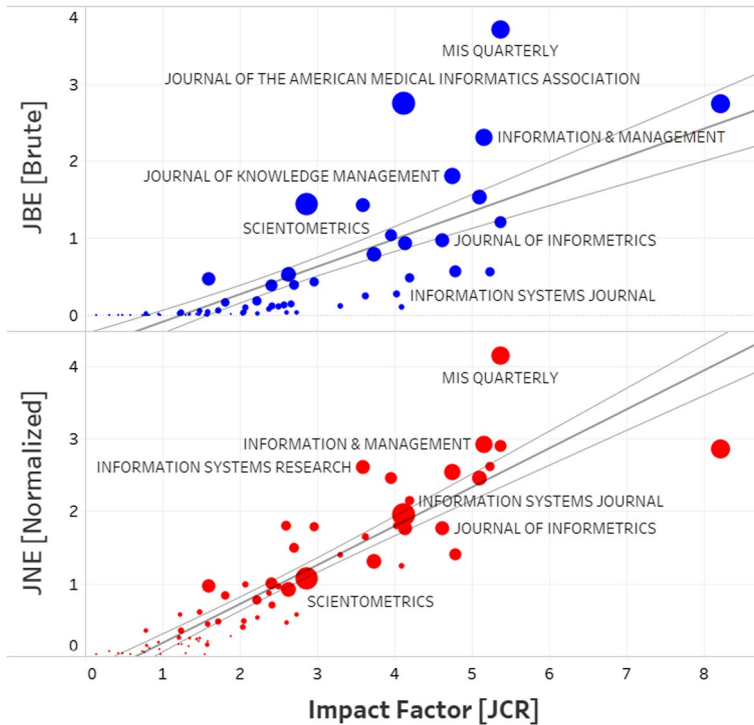


Fig. 1 Correlations of the JIF with the JBE and JNE in “Information Science & Library Science” (ILS)

Table 6 Pearson correlations between all measures and indicators for all JCR journals in “Information Science & Library Science” (ILS) (lower left triangle: articles and reviews; upper right triangle: all document types; published between 2009 and 2018)

Art. + Rev. vs. all	Pubs	Exc. Pubs	JIF	JPEP	JCEC	JBE	JNE
Pubs	–	0.077	–0.136	–0.107	0.077	–0.107	–0.029
Exc. Pubs	0.775	–	0.683	0.696	1.000	0.696	0.890
JIF	0.253	0.700	–	0.897	0.683	0.897	0.755
JPEP	0.222	0.716	0.906	–	0.696	1.000	0.830
JCEC	0.775	1.000	0.700	0.716	–	0.696	0.890
JBE	0.462	0.890	0.763	0.836	0.890	–	0.830
JNE	0.222	0.716	0.906	1.000	0.716	0.836	–

and size-independent indicator (JNE) is used (e.g., *Journal of the Association for Information Science and Technology* and *Scientometrics*). Table 6 shows the Pearson correlations between all five indicators (JPEP, JCEC, JIF, JBE and JNE) for the category “Information Science & Library Science” (ILS) for (a) only articles and reviews (lower left triangle), and (b) for all the document types (upper right triangle).

In Table 7, journals in the category “Information Science & Library Science” (ILS) are listed. It shows the changes in ranking position, which is traditionally based on the JIF, when applying the excellence indicators JBE and JNE. *Portal: Libraries and the Academy* and *Journal of Health Communication* are the journals that improve their rank position the

Table 7 Ranking changes for journals in “Information Science & Library Science” (ILS) according to the JIF in comparison with JBE and/or JNE (article and reviews, published between 2009 and 2018)

	Ranks			Differences between rankings		
	JIF	Brute excellence JBE	Normalized excellence JNE	JIF and JBE	JIF and JNE	
Scientometrics	21	7	23	+14	-2	
Electronic Library	64	51	58	+13	+6	
Online Information Review	39	26	32	+13	+7	
Library & Information Science Research	47	35	35	+12	+12	
Library Hi Tech	57	46	47	+11	+10	
European Journal of Information Systems	27	17	13	+10	+14	
Information Systems Research	18	8	6	+10	+12	
Journal of the American Medical Informatics Association	12	2	11	+10	+1	
Ethics and Information Technology	35	33	25	+2	+10	
Journal of Informetrics	9	11	15	-2	-6	
Professional de la Informacion	44	49	57	-5	-13	
Knowledge Organization	59	68	70	-9	-11	
Library Collections Acquisitions & Technical Services	52	62	60	-10	-8	
ASLIB Journal of Information Management	34	44	38	-10	-4	
Information and Organization	19	30	20	-11	-1	
Journal of Strategic Information Systems	4	15	5	-11	-1	
Revista Espanola de Documentacion cientifica	53	66	68	-13	-15	

Completed Table available at conference paper: <https://doi.org/10.5281/zenodo.5679387>

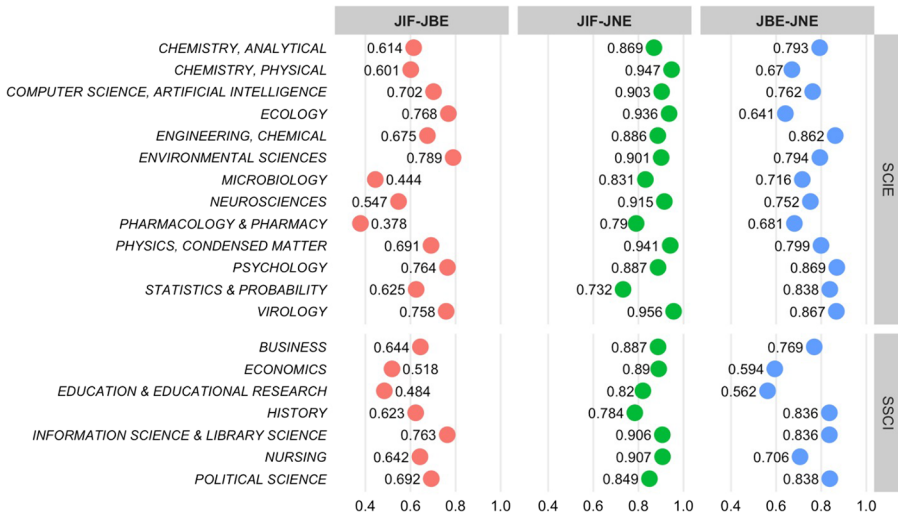


Fig. 2 Correlations between JIF, JBE and JNE for the 20 subject categories analyzed (article and reviews, published between 2009 and 2018)

most due to the excellence indicators. *Malaysian Journal of Library & Information Science* and *Information Technology for Development* are the ones decreasing the most in the brute and normalized excellence rankings.

Comparisons between the 20 categories analyzed

Figure 2 shows the results of the correlation between the JIF and the two excellence indicators for all 20 categories considered in our study. The results show that the correlation between the JIF and the JNE is higher than between JIF and the JBE. This is expected because JIF and JNE are both size independent. The highest correlation between the JIF and the JNE is in the JCR category “Virology” (VIR), followed by “Chemistry, Physical” (CHPH), “Physics, Condensed Matter” (PHCM), “Ecology” (ECOL), “Neurosciences” (NEUR), “Nursing” (NURS) and “Computer Science, Artificial Intelligence” (COMP), (all of them above 0.9). The lowest one is reported by “Statistics & Probability” (STPR), followed by “History” (HIS) and “Pharmacology and Pharmacy” (PHAR) (all of them under 0.8). The correlations between the JIF and the JBE are only moderate and much lower than the ones between JIF and JNE. The lowest is reported by the category “Pharmacology and Pharmacy” (PHAR) (0.378) and the highest by “Environmental Sciences” (ENV) (0.789). These results reveal that the journal excellence content is not completely reflected in the JIF measure, and this affects both Science (SCIE) and Social Science Edition (SSCI) categories, as it can be seen very clearly in Fig. 2.

Effect of the multidisciplinary

Table 8 lists the journals and categories where multidisciplinary journals have the highest contribution to the indicator JCEC and only journals with greater than 0.5 have been included. Results illustrates strong differences in the effects of the “multidisciplinary

Table 8 Effect of the multidisciplinary journals in the JBE ranking for the journals of the 20 subject categories (article and reviews, published between 2009 and 2018)

Cat	Journal	Journal contribution to the excellence of the category (JCEC)**	Journal percentage of excellent publications (JPEP) (%)	JBE Brute excellence	JBE rank
CHPH	Nature Communications	0.528	47	0.248	19/161
ECOL	PNAS	3.552	64	2.290	3/168
ECOL	PLOS ONE	3.417	8	0.283	24/168
ECOL	Science	1.408	88	1.244	5/168
ECOL	Scientific Reports	1.243	13	0.167	33/168
ECOL	Nature	1.157	94	1.082	6/168
ECOL	Nature Communications	1.102	53	0.586	11/168
ENV	PNAS	0.504	73	0.370	16/274
ILS	PLOS ONE	1.311	39	0.514	17/86
MICR	PLOS ONE	4.860	13	0.633	15/168
MICR	PNAS	3.029	65	1.955	5/168
MICR	Scientific Reports	2.036	17	0.347	21/168
MICR	Nature Communications	1.129	67	0.757	11/168
MICR	Nature	0.710	96	0.685	13/168
MICR	Science	0.689	90	0.622	16/168
NEUR	PNAS	4.386	44	1.925	4/299
NEUR	PLOS ONE	2.696	7	0.193	23/299
NEUR	Nature Communications	1.505	42	0.636	13/299
NEUR	Nature	1.492	95	1.420	6/299
NEUR	Science	1.341	87	1.166	7/299
NEUR	Scientific Reports	1.309	9	0.118	28/299
PHAR	PLOS ONE	0.652	15	0.098	44/323
STPR	PLOS ONE	0.521	12	0.062	38/150
VIR	PNAS	4.474	49	2.210	4/50

Table 8 (continued)

Cat	Journal	Journal contribution to the excellence of the category (JCEC)**	Journal percentage of excellent publications (JPEP) (%)	JBE Brute excellence	JBE rank
VIR	PLOS ONE	2.258	5	0.111	13/50
VIR	Science	1.617	98	1.590	5/50
VIR	Nature Communications	1.436	59	0.840	7/50
VIR	Nature	1.352	97	1.312	6/50
VIR	Scientific Reports	1.185	9	0.103	14/50

**Only values of Journal Contribution to the Excellence of the Category (JCEC) greater than 0.5 have been included

journals” in the 20 selected subject categories. Categories related to the life sciences and natural sciences show the strongest influences of such journals compared with the Social Science that are less affected or where these journals are not representative. We have to keep in mind that humanities and most fields in the social sciences have a lesser weight in the big multidisciplinary journals. In more than half of the categories studied at least one or more multidisciplinary journals would be Q1 if they will be considered in JCR as part of the category. In particular, “Ecology” (ECOL), “Virology” (VIR), “Neurosciences” (NEUR) and “Microbiology” (MICR) are the categories with the highest presence of multidisciplinary journals. In these categories, six multidisciplinary journals are responsible for a very large brute excellence contribution and can be considered as “Q1 journals” in this category according to this indicator.

If we consider our indicators of excellence, *Proceedings of the National Academy of Sciences (PNAS)* would be one of the most important multidisciplinary journals. *PNAS* ranks 3rd in “Ecology” (ECOL) and 4th respectively in “Virology” (VIR) and “Neurosciences” (NEUR). The journal *Science* also stands out in 5th position in “Ecology” (ECOL) and “Virology” (VIR). Open access journals or megajournals also stand out as journals that contribute excellent papers to the categories. We refer to three journals: *Nature Communications*, *Scientific Reports* and *PLOS ONE*. The latter stands out for being present in almost all categories with a substantial contribution of excellent papers. The only multidisciplinary journal ascending to the first quartile in “Information Science & Library Science” (ILS) is *PLOS ONE*. As it is well-known, *PLOS ONE* has a special section for Research assessment and Bibliometrics. However, according to its size, its excellence contribution is not as high as expected.

Effect of the document types

Finally, we analyzed the effect of considering all types of documents instead of only articles and reviews. As it is common knowledge that there is an asymmetry in the calculation of the JIF. In the numerator, the citations to all types of documents are summed up, while in the denominator only research articles and reviews are considered.³ In “[Introduction](#)” we have already analyzed the document types in each category and their contribution to the excellence (see [Table 1](#)). The results corroborate that in the subject categories related to the social sciences, e.g., “Information Science & Library Science” (ILS) and “History” (HIS), other document types than articles and reviews might play a significant role accounting for around 18% of the category excellence. Furthermore, the two new excellence indicators have been also calculated for all document types and for articles and reviews only (see [Table 3](#)). The results underline the role of research articles and reviews in scientific journals. Any reasonable correlation of the number of documents with excellence measure is absent, even slightly negative. Thus, it is plausible that the observed Pearson correlation between JIF and JNE is distinctly higher for articles and reviews than for all document types (0.906 versus 0.755), while it is just the opposite for the brute excellence contribution (JBE), where the total number of publications in the category plays a role (0.763 versus 0.897).

³ Originally, Garfield used the document types, articles and reviews, also called “citable items” in the JCR Edition. Nowadays, all the proceedings papers published in journals are also considered articles in the WoS Core Collection with the effect of double assignment.

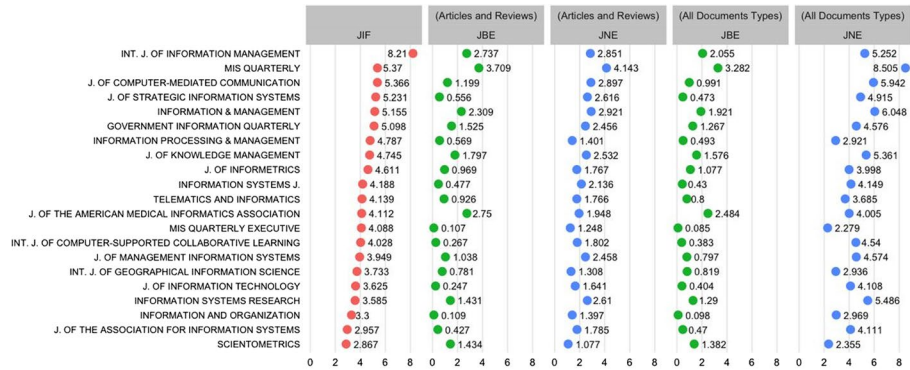


Fig. 3 Distribution of JIF, JBE and JNE for all Q1 journals in “Information Science & Library Science” (ILS) for only articles and reviews (2nd and 3rd columns 2 and 3) and all document types (columns 4 and 5)

Figure 3 shows the correlation of the Journal Impact Factor, and the two excellence indicators (JBE and JNE) for the Q1 journals of the category in “Information Science & Library Science” (ILS) when considering only articles and reviews (column 2 and 3) and all document types (column 4 and 5), respectively. The results show that, even if the actual indicator values are changing, the distribution of the JBE or JNE as such is not much affected by the considering all document types instead of only “citable items”. This hints to the fact that our excellence indicators are quite robust or less sensitive to the types of documents considered. In particular, the correlations are very strong, e.g., 0.986 for the JNE, and 0.99 for the JBE, and they corroborate the robustness of both indicators concerning the document types used in their calculation. One possible reason is that the normalizations performed for defining excellent publications are also done by document type (Top 10% most cited publications of the same document type and publication year in the same category year).

Final remarks

Due to the precariousness and long half-lives of the citations, the identification of the top journals in each discipline is one of the most requested and used tools in academic evaluation exercises focusing on the assessment of the research performance of the most recent years. Despite the enormous criticism it has received in scientific articles and manifestos, the JIF has established itself as one of the most consolidated instruments for assessing the impact and prestige of the journals where the scientists, research groups, organizations and countries have published in. To provide a broader view of each journal’s contribution to the excellence in each category or field, we have introduced two new indicators, which ideally complement the JIF. The first one, the Journal Normalized Excellence (JNE) measures the normalized excellence contribution of a journal to its subject category. A journal contributes relatively more (less) to the subject’s excellence according as $JNE > (<) 1$. On the other side, it is also interesting to know the total contribution of a journal to the category excellence, independently of its size. The Journal Brute Excellence (JBE) reflects the total brute excellence force or brute contribution of the journal to the category. Similarly, in

those cases in which a journal is present in several categories, the JBE allows a better comparison of its performance in each of them.

The case study applied to “Library Science & Information Science” (ILS) has shown how the indicators work. It has been observed how JBE has allowed us to identify which journals contribute the most significant, i.e., excellent, papers to a journal. Likewise, the JNE indicator has allowed us to contextualize the Gross Contribution with the total number of documents in the journal. In this sense, the proposal yields positive results, firstly because they provide different and complementary information, as demonstrated by their different correlation with the Journal Impact Factor. The correlations are similar in almost all the categories analyzed. Thus, the correlation is moderate between JIF-JBE and high/significant between JIF and JNE, with the exception of singular cases such as “Statistics & Probability” (STPR). Among the proposed indicators, JBE and JNE, the correlation was moderate/low with cases of no correlation (e.g., “Education (EDU) and “Economics” (ECO)). This situation is interesting as we are dealing with two different and complementary indicators but with singularities according to the categories although less pronounced than in the JIF (Dorta-Gonzalez & Dorta-González, 2013). On the other hand, although percentile-based indicators may have limitations (Bornmann et al., 2013), the correlations of the JIF with JNE indicate that no information is lost. It also overcomes certain limitations of the JIF such as the Citation Window. Although it is well known that journal impact measures do not work well in the Arts and Humanities and can lead to false interpretations (Repiso et al., 2019). In this sense, for the two indicators in “History” (HIS), the results are comparable to those of the other scientific categories.

In this study, including 20 WoS subject categories, our excellence indicators have shown a robustness concerning the consideration of all types of documents instead of only articles and reviews. Therefore, they provide an amelioration of the inherent asymmetry reflected in the definition and calculation of Garfield’s Impact Factor. Another advantage of our excellence indicators relies on the practical aspect for the measurement of the visibility of publications. When using the JIF for this purpose, there is always a controversial decision: what JCR Edition should be used? There are three possibilities: (a) using JIF values of the last JCR-edition for all publications independently of their publication year; (b) Using the JCR-edition corresponding to the publication year of each publication; and (c) using the mean value of the last x years according to the time period under study. Any of them is completely satisfactory (Glänzel et al., 2016). Excellence indicators circumvent this problem because they are based on accumulated measures including the last ten complete publication years⁴ and are not restricted to 2 years or a selected JCR edition.

This study also revealed that the effect of the multidisciplinary journals differs according to the category, and this effect is generally stronger in the so-called ‘hard sciences. One of the possible applications of our study is to prevent the use of JCR categories for the delineation of scientific areas, as has been done in many previous bibliometric studies. The study warns of serious consequences of this approach, as contributions from multidisciplinary journals are not considered in some categories. For example, reducing the study to only journals of the category in “Ecology” (ECOL), “Statistics & Probability” (STPR) or “Microbiology” (MICR) would mean missing a large part of the scientific breakthroughs and excellent publications, which are regularly published in multidisciplinary journals specially *PNAS*, *Science*, *Nature* and *PLOS ONE*.

⁴ The number of years considered in the calculation could also be variable and depending on the cited half-lives of each category.

In relation to *PLOS ONE* our study agrees with previous results, which show the multidisciplinary nature of this journal and how in certain JCR categories it has a significant impact (Repiso et al., 2020).

Another interesting question is the effect of interdisciplinarity. Unfortunately, InCites does not offer the possibility to measure this effect, because the subject classification is made on journal level, except for the multidisciplinary journals (on publication level). The recent introduction of the publication based “Citation Topics” may be an improvement in InCites. This topic will also be part of our future analyses.

In the most recent edition of the JCR, a new indicator has been introduced, the JCI, which is based on the CNCI. This indicator is mainly used for Arts & Humanities and ESCI journals, for which the journal impact factor is not calculated, following Garfield’s recommendations. However, this indicator is also a mean value (like the JIF) and does not consider the skewness of the distribution of citations (Bornmann et al., 2013). Therefore, single outliers, extremely highly cited papers, can distort dramatically his values (Antonoyiannakis, 2019; Dimitrov et al., 2010). The use of the excellence indicators as suggested in this study will provide a much better assessment of the impact of the journals.

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Data availability All data are available at <https://doi.org/10.5281/zenodo.5676184>.

Declarations

Conflict of interest Wolfgang Glänzel is editor-in-chief of *Scientometrics*.

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