

Visualizing Music Psychology: A Bibliometric Analysis of Psychology of Music, Music Perception, and *Musicae Scientiae* from 1973 to 2017

Music & Science

Volume 2: 1–18

© The Author(s) 2019

Article reuse guidelines:

sagepub.com/journals-permissions

DOI: 10.1177/2059204318811786

journals.sagepub.com/home/mns



Manuel Anglada-Tort¹  and Katie Rose M. Sanfilippo²

Abstract

Music psychology has grown drastically since being established in the middle of the 19th century. However, until now, no large-scale computational bibliometric analysis of the scientific literature in music psychology has been carried out. This study aims to analyze all published literature from the journals *Psychology of Music*, *Music Perception*, and *Musicae Scientiae*. The retrieved literature comprised a total of 2,089 peer-reviewed articles, 2,632 authors, and 49 countries. Visualization and bibliometric techniques were used to investigate the growth of publications, citation analysis, author and country productivity, collaborations, and research trends. From 1973 to 2017, with a total growth rate of 11%, there is a clear increase in music psychology research (i.e., number of publications, authors, and collaborations), consistent with the general growth observed in science. The retrieved documents received a total of 33,771 citations ($M = 16.17$, $SD = 26.93$), with a median (Q1—Q3) of 7 (2—20). Different bibliometric indicators defined the most relevant authors, countries, and keywords as well as how they relate and collaborate with each other. Differences between the three journals are also discussed. This type of analysis, not without its limitations², can help understand music psychology and identify future directions within the field.

Keywords

Bibliometrics, music psychology, psychology of music, scientometrics, visualization technique

Introduction

The beginnings of what we now regard as music psychology started in the middle of the 19th century as a branch of both psychology and musicology (Thaut, 2011). But music psychology has evolved and grown drastically since then. From a focus on psychoacoustics, perception, and the cognitive sciences, to health applications and the use of music in everyday life, music psychology has shifted and blossomed, establishing programs, labs and journals covering different research interests, geographical areas, and research groups.

Music psychology can be defined as the scientific study of the psychological processes through which music is perceived, created, responded to, and incorporated into everyday life (Tan, Pfordresher, & Harré, 2017; Thompson, 2009). The field of music psychology therefore embraces an incredibly diverse and wide variety of topics, including the origins of music, music perception and cognition, responses to music (e.g., bodily, emotional, and aesthetic),

the neuroscience of music, music development, music education, music performance, composition and improvisation, the use of music in everyday life, and music therapy and wellbeing (Hallam, Cross, & Thaut, 2011). But the psychology of music can also contribute to broader fields, such as social psychology, behavioral science, aesthetics, computer science, medicine and health, consumer psychology, marketing, and advertising. Researchers from all over the globe investigate these topics empirically, with more than 80 music cognition and science labs around the world (www.musicperception.org/smpc-resources.html). Psychology of music conferences have been held in the UK

¹ Technische Universität Berlin, Berlin, Germany

² Goldsmiths College, University of London, London, UK

Corresponding author:

Manuel Anglada-Tort, Department of Audio Communication, Technische Universität Berlin, Berlin, Germany.

Email: m.angladatort@campus.tu-berlin.de



Creative Commons Non Commercial CC BY-NC: This article is distributed under the terms of the Creative Commons Attribution-NonCommercial 4.0 License (<http://www.creativecommons.org/licenses/by-nc/4.0/>) which permits non-commercial use, reproduction and distribution of the work without further permission provided the original work is attributed as specified on

the SAGE and Open Access pages (<https://us.sagepub.com/en-us/nam/open-access-at-sage>).

since 1972 under the umbrella of SEMPRES (formerly SRPMME). Since then, various music-psychology-specific conference series have also begun to develop, such as the International Conference on Music Perception and Cognition (ICMPC), founded in 1989, and the European Society for the Cognitive Sciences of Music (ESCOM), founded in 1991. In 2008, the International Conference of Students of Systematic Musicology (SysMus) was founded for students of systematic musicology, a broader field which encompasses music psychology.

The first research journal specifically dedicated to music psychology is *Psychology of Music*, established in 1973. This multidisciplinary journal's aim is to "increase scientific understanding of all psychological aspects of music and music education" (journals.sagepub.com/home/pom). *Music Perception*, established in 1983, was developed with a primary focus on cognitive-psychological research with broader and multidisciplinary draw, including work from "psychology, psychophysics, neuroscience, music theory, acoustics, artificial intelligence, linguistics, philosophy, anthropology and cognitive science" (mp.ucpress.edu). In 1997, the European Society for the Cognitive Sciences of Music (ESCOM) was developed along with its journal *Musicae Scientiae*, which aims to include "empirical, theoretical and critical articles directed at increasing understanding of how music is perceived, represented and generated" (journals.sagepub.com/home/msx). As a truly multidisciplinary subject, music psychology research is published in many other journals, including other APA journals and journals from related disciplines, such as musicology, music theory, music therapy, music education, aesthetics, marketing, and neuroscience. This includes, for example, the *Journal of Research in Music Education*, *International Journal of Music Education*, *Journal of Music Therapy*, *Empirical Musicology Review*, and *Psychomusicology: Music, Mind, and Brain*.

The current research focuses on the three most prominent scientific journals in music psychology, namely *Psychology of Music*, *Music Perception*, and *Musicae Scientiae*. We used two criteria to select these journals: content and impact. Regarding content, the focus was on journals covering specifically the psychology of music. Impact was determined by the SJR ranking provided in SCImago (www.scimagojr.com). This measure indicates the average number of weighted citations per document received within the selected journal during the previous three years. In June 2018, searching in the category "music," *Psychology of Music* was ranked fourth, *Musicae Scientiae* sixth, and *Music Perception* seventh. The first (*IEE Signal Processing Magazine*), second (*Journal of Research in Music Education*), third (*Music Education Research*), and fifth (*International Journal of Music Education*) journals did not meet the first criterion of content, focusing on other topics rather than on music psychology (i.e., signal processing or music education).

With the surge of interest in music psychology research, it is important as a discipline to reflect systematically on what research has been published and what gaps can still be filled. Bibliometrics and scientometrics allow for the measurement and analysis of published scientific literature, giving objective and measurable data to help us understand the discipline's trajectory thus far. By using computational, mathematical and statistical techniques, bibliometrics analyses the quantity and quality of published scientific literature, including citation analysis, authorship and country productivity and collaborations, impact of publications, and research trends (e.g., Blázquez-Ruiz, Guerrero-Bote, & Moya-Anegón, 2016; Blažun, Kokol, & Vošner, 2015; Chen, Arsenault, Gingras, & Larivière, 2015; De Bellis, 2009; Laengle et al., 2017; Mryglod, Holovatch, Kenna, & Berche, 2016; Naukkarinen & Bragge, 2016; Sweileh, 2017; Sweileh et al., 2016, 2017; Sweileh, Al-Jabi, Sawalha, & Zyoud, 2016).

Bibliometric analyses have rarely been applied to music psychology literature. We have found two articles that used a bibliometric approach to study perception and cognition research (Tirovolas & Levitin, 2011) and music and affect research (Diaz & Silveria, 2014). In Tirovolas & Levitin's (2011) study, the authors looked at publications within one journal (*Music Perception*), covering a total of 578 articles. The retrieved literature was coded to look at the most frequent topics, populations, stimuli, materials, outcome measures, and music styles, indicating their trends between 1984 and 2009. They also provided a list of the top 20 most highly cited articles published in *Music Perception* and the top 20 articles published outside *Music Perception* that were most cited in the journal. Finally, the authors showed the most productive countries publishing in *Music Perception*. In the paper by Diaz and Silveria (2014), the authors looked specifically at music and affective phenomena. They focused on three journals: the *Journal of Research in Music Education*, *Psychology of Music*, and *Music Perception*. The authors used strict inclusion criteria to select articles related to topics relevant to affective aspects of music, resulting in a total of 286 articles.

A recent paper by Sloboda and Ginsborg (2018) investigated the country and discipline spread of the contributors and members of *Musicae Scientiae* as well as compared the topics of early meetings and publications with current meetings. Although a strict bibliometric approach was not used, the paper is the only publication to specifically examine the international spread of music psychology researchers. While the studies by Tirovolas & Levitin (2011) and Sloboda and Ginsborg (2018) only focused on one scientific journal, the study by Diaz and Silveria (2014) had a very narrow topical focus (i.e., music and affect research). Thus, these studies cannot give insight into trends throughout music psychology as a whole. Moreover, the studies used very few bibliometric indicators. For instance, they did not provide information about the growth of publications, more elaborated citation analysis, or author

productivity and collaborations. Other important limitations in these studies are the relatively small datasets and the use of human coders to analyze the content of the articles.

The present study aims to produce a large-scale computational bibliometric analysis of the scientific literature published in music psychology from 1973 to 2017. Using this method of analysis we aim to better understand research trends, citations, authorships, collaborations, as well as global contributions. This is important in identifying future directions within the field. To reduce potential sources of bias and analyze systematically a large amount of documents, the present study used the R package *bibliometrix* (Aria & Cuccurullo, 2017), a tool for quantitative research in bibliometrics that provides various functions to perform citation, coupling, and scientific collaboration analysis. To visualize the data, we used *VOSviewer* (Van Eck & Waltman, 2010), a software tool that applies advanced clustering and natural language processing techniques for generating and visualizing maps based on network data. *VOSviewer* software has been used in a large body of published literature (www.vosviewer.com/publications), generating over 500 publications since 2006. To the best of our knowledge, the software has not yet been applied to music psychology literature.

In the present study we analyze, through visualization and bibliometric techniques, all published literature from *Psychology of Music*, *Music Perception*, and *Musicae Scientiae*, while focusing on five key aspects of the retrieved literature: (1) growth of publications (i.e., annual growth rate, relative growth rate, and whether there are significant temporal changes in the number of publications over time), (2) citation analysis (i.e., number of citations per journal and year, top cited authors and papers, and whether there are significant temporal changes in the number of citations over time), (3) authorship analysis (i.e., productivity, dominance, collaboration index, visualizations of authorship collaboration, and Lotka's law coefficient for scientific productivity), (4) country analysis (i.e., productivity, visualization of country collaboration, and geographical distribution of the publications), and (5) the main conceptual language used in the retrieved literature.

Methods

Data collection and search strategy

The data used in this study was retrieved from Scopus, a bibliographic database that covers over 20,000 journals, including technical, medical, and social sciences titles. Scopus is larger than PubMed and Web of Science (Falagas, Pitsouni, Malietzis, & Pappas, 2008) and offers many relevant features that facilitate bibliometric analysis (e.g., author, country, and affiliation contributions, citation analysis, and the "source type" function).

We searched all available literature, by "source title", in *Psychology of Music*, *Music Perception*, and *Musicae Scientiae*. Using the Scopus "source type" function, we limited the search to empirical and review articles only, excluding book chapters, conference papers, and editorial notes. We also excluded any document from 2018 because it was the year in which this study was conducted. All available results were then exported to text files, including citation information (i.e., authors, document title, year, source title, volume, issue, pages, citation count, source and document type, and DOI), bibliographical information (i.e., affiliations, serial identifiers, publisher, editor, language of original document, correspondence address, and abbreviated source title), abstracts, as well as keywords. All data was retrieved on April 20, 2018 (see supplementary materials for the two main datasets used in this study).

In some situations, the same author might have more than one name, use different initials in different publications (e.g., Sloboda, J. vs. Sloboda, J. A.), or have different name spellings. This might generate inaccuracy and inconsistencies in the computational analysis of authorship. There is not a general solution to this problem. Researchers can decide between two imperfect approaches: (a) to analyze the data without any previous processing, which would count the same authors as different authors when their names were spelled using different initials; or (b) to remove the initials of the second (and third) names from all authors, which would count two different authors as the same when they had the same surname and first initial. We chose the latter approach because it was considered to have a smaller negative impact on the analysis. Moreover, we did not find any case in which two different authors shared the same surname and first initial names in our dataset. Thus, we removed the second and third initial in all authors, including only the first surname and first initial.

In addition, it is important to note that our data has a gap in the literature retrieved from *Music Perception* between 2002 and 2004, as Scopus does not contain any documents from this source during these three years. Although *Music Perception* published articles during this period of time, we do not have a clear reason to explain this gap.

Data analysis and visualization

Descriptive statistics and standard bibliometric indicators, including citation analysis, annual growth of publications, authorship productivity, dominance, collaboration index, and country productivity were used to produce an overview of the retrieved data. The application and presentation of some of these indicators was based on the analysis reported in Sweileh et al. (2017). In addition, we used the R package *bibliometrix* (Aria & Cuccurullo, 2017) to analyze the most productive authors, countries, keywords, top cited articles and authors, author dominance, index-h, and Lotka's law coefficient for scientific productivity (1926).

Visualization and bibliometric maps were created using *VOSviewer* (Van Eck & Waltman, 2010), which uses a

Table 1. Number and type of articles retrieved.

Type of document	PoM (1973–2017)	MP (1983–2017)	MS (1997–2017)	Total	%
Research articles	872	710	405	1,987	95.12
Review articles	62	36	4	102	4.88
Total	934	746	409	2,089	100
%	44.71	35.71	19.58	100	–

Note. PoM: *Psychology of Music*; MP: *Music Perception*; MS: *Musicae Scientiae*.

Table 2. Top 20 contributions of authors, keywords, and countries.

Author	TP	Keywords	TP	Country*	TP
Repp, B. H.	27	Music	183	U.S.	484
Kopiez, R.	22	Emotion	87	UK	330
Schubert, E.	22	Performance	45	Canada	130
Huron, D.	21	Music perf.	39	Australia	118
Sloboda, J. A.	21	Rhythm	37	Germany	102
Davidson, J. W.	20	Music training	33	Finland	62
MacDonald, R. A. R.	20	Perception	33	France	56
North, A. C.	19	Music therapy	29	Netherlands	47
Clarke, E. F.	17	Creativity	28	Japan	41
Eerola, T.	17	Singing	28	Israel	33
Toiviainen, P.	17	Motivation	25	Sweden	31
Trehub, S. E.	17	Memory	24	Belgium	29
Hargreaves, D. J.	16	Music listening	24	Austria	22
Thompson, W. F.	16	Music perception	24	Italy	21
Welch, G. F.	16	Personality	24	Spain	15
Krumhansl, C. L.	15	Practice	24	Georgia	13
Müllensiefen, D.	14	Preference	24	Poland	9
Williamson, A.	14	Arousal	23	Switzerland	9
Cross, I.	13	Expertise	23	Norway	8
Hallam, S.	13	Communication	22	Greece	7

Note. TP: total publications. *Country of corresponding author.

unified framework for mapping and clustering (Waltman, Van Eck, & Noyos, 2010). The software is mainly intended for analysis of bibliometric networks and can create three types of visualizations: network visualizations, overlay visualizations, and density visualizations. In the network visualizations, items are represented by their label and by a circle. The size of the circles is determined by the weight of the item. The place of the items in the map and their colors are used to cluster the items. The color of an item is determined by the cluster to which the item belongs. Lines between items represent links and the stronger the link is, the wider the line. The distance between items in the map indicates the degree of relatedness between them. Furthermore, we used the R package *rworldmap* (South, 2011) to generate visualizations of the geographical distribution of countries' productivity.

Results

Retrieved literature

A total of 2,089 documents were retrieved, covering a time period of 44 years (1973–2017) beginning from the

first publication of *Psychology of Music* in 1973. Table 1 shows the total number and type of articles retrieved per journal and in total. The majority of documents were research articles (1,987; 95.12%), whereas review articles only represented a minimal portion (102; 4.88%). *Psychology of Music* was the journal with the largest number of retrieved articles (934; 44.71%), followed by *Music Perception* (746; 35.71%), and *Musicae Scientiae* (409; 19.58%). However, when taking into account the years that each journal has been active, the average number of publications per year is comparable across the three journals (20.76, 23.31, and 19.48, respectively). Table 2 shows the top 20 contributions made by author, keywords, and countries. See Appendix A for the tables of the top 20 contributions made by author, keyword, and country by decade, and Appendix B for the top 10 contributions by each journal.

Growth in number of publications

The mean number of publications from 1973 to 2017 was 46.42 ($SD = 35.56$). The total percentage of relative growth was 11%. The highest productivity was observed in 2016

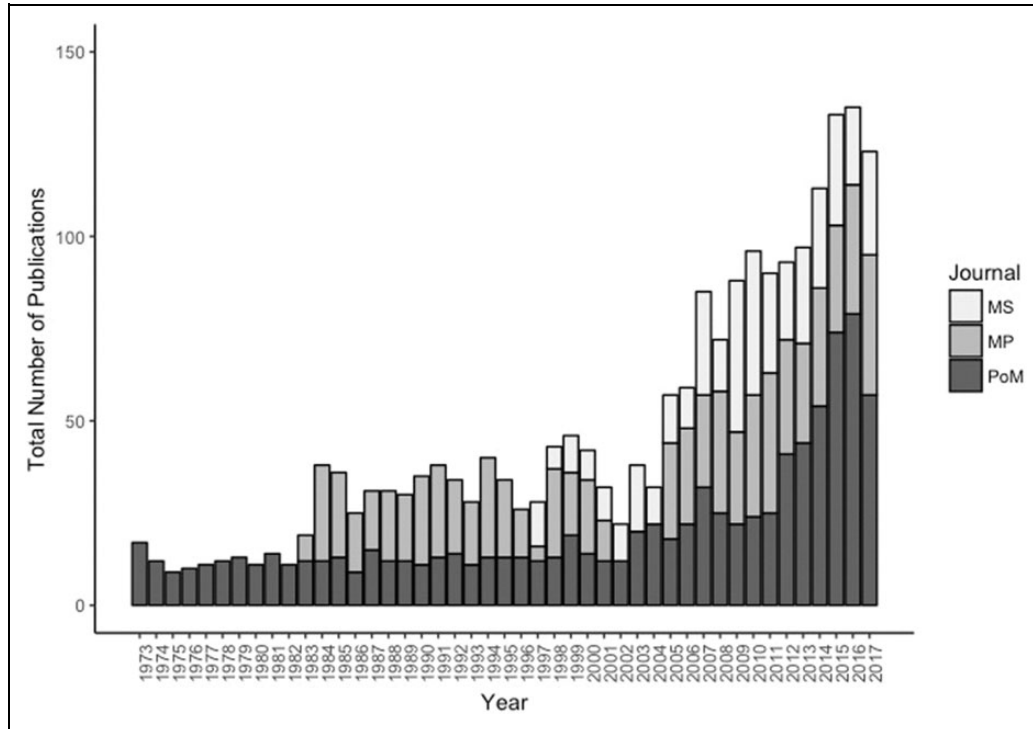


Figure 1. Total number of publications per journal over time. Note: PoM: *Psychology of Music*; MP: *Music Perception*; MS: *Musicae Scientiae*.

with a total of 135 publications (6.46%) and the lowest productivity was observed in 1975 with a total of nine publications (.43%). Figure 1 shows the total number of publications in the three journals over time. The total number of publications increased significantly over time, as indicated by a simple linear regression, $F(1,43) = 141.1$, $p < .001$, with an R^2 of .766.

Table 3 shows the annual number of publications, annual growth rate (AGR), and relative growth rate (RGR). The AGR indicates the percentage of change in the number of publications over one year. The AGR is calculated using the following equation: $AGR = [(TP \text{ ending value} - TP \text{ beginning value}) / TP \text{ beginning value}] * 100$, where TP is total number of publications. The RGR indicates the growth rate relative to the total number of publications per year. The RGR was calculated based on the following equation: $RGR = [\log_e W_2 - \log_e W_1] / (T2 - T1)$, where $\log_e W_2$ is the log of the final number of publications after a specific period of interval; $\log_e W_1$ is the log of the initial number of publications; and $T1 - T2$ is the unit difference between the initial time and the final time.

Appendix C shows the annual number of publications, AGR, and RGR in the three journals separately. In *Psychology of Music*, the average number of publications from 1973 to 2017 was 20.76 ($SD = 16.48$), with a total relative growth rate of 9%. In *Music Perception* the mean number of publications from 1983 to 2017 was 23.31 ($SD = 8.31$), with a total relative growth rate of 15%. In *Musicae Scientiae*, the average number of

publications from 1997 to 2017 was 19.48 ($SD = 10.38$), with a total relative growth rate of 18%.

Citation analysis

Table 4 shows the summary of the citation analysis of all three journals combined. Retrieved documents received a total of 33,771 citations, a mean of 16.17 ($SD = 26.93$) citations per document, and median (Q1–Q3) of 7 (2–20). While the highest number of total citations was in 2007, with 1,978 ($M = 23.3$, $SD = 30.1$) citations, the lowest was in 1975, with 25 citations ($M = 2.8$, $SD = 2.9$). Figure 2 shows the average total number of citations over time. Across the entire time period, the average number of citations did not increase significantly, as indicated by a simple linear regression, $F(1,43) = .21$, $p = .65$, $R^2 = .005$. However, the relationship between the average citations and year followed an inverted-U shape, as indicated by a statistically significant quadratic regression, $F(2,42) = 52.65$, $p < .001$, $R^2 = .715$.

Appendix D shows the summary of citation analysis in the three journals separately. In *Psychology of Music*, the retrieved documents received a total of 13,344 citations, a mean of 16.98 ($SD = 26.12$) citations per document, and median (Q1–Q3) of 8 (3–21). In *Music Perception*, the documents received a total of 17,069 citations, a mean of 24.38 ($SD = 33.25$) citations per document, and median (Q1–Q3) of 14 (5–29). In *Musicae Scientiae*, the

Table 3. Annual number of publications, AGR, and RGR.

Year	Frequency (%)	AGR	Cumulative total	Log _e W	RGR
1973	17 (.81)	–	17	2.83	–
1974	12 (.57)	–.29	29	3.37	.53
1975	9 (.43)	–.25	38	3.64	.27
1976	10 (.48)	.11	48	3.87	.23
1977	11 (.53)	.10	59	4.08	.21
1978	12 (.57)	.09	71	4.26	.19
1979	13 (.62)	.08	84	4.43	.17
1980	11 (.53)	–.15	95	4.55	.12
1981	14 (.67)	.27	109	4.69	.14
1982	11 (.53)	–.21	120	4.79	.10
1983	19 (.91)	.73	139	4.93	.15
1984	38 (1.82)	1.00	177	5.18	.24
1985	36 (1.72)	–.05	213	5.36	.19
1986	25 (1.2)	–.31	238	5.47	.11
1987	31 (1.48)	.24	269	5.59	.12
1988	31 (1.48)	.00	300	5.70	.11
1989	30 (1.44)	–.03	330	5.80	.10
1990	35 (1.68)	.17	365	5.90	.10
1991	38 (1.82)	.09	403	6.00	.10
1992	34 (1.63)	–.11	437	6.08	.08
1993	28 (1.34)	–.18	465	6.14	.06
1994	40 (1.91)	.43	505	6.22	.08
1995	34 (1.63)	–.15	539	6.29	.07
1996	26 (1.24)	–.24	565	6.34	.05
1997	28 (1.34)	.08	593	6.39	.05
1998	43 (2.06)	.54	636	6.46	.07
1999	46 (2.2)	.07	682	6.53	.07
2000	42 (2.01)	–.09	724	6.58	.06
2001	32 (1.53)	–.24	756	6.63	.04
2002	22 (1.05)	–.31	778	6.66	.03
2003	38 (1.82)	.73	816	6.70	.05
2004	32 (1.53)	–.16	848	6.74	.04
2005	57 (2.73)	.78	905	6.81	.07
2006	59 (2.82)	.04	964	6.87	.06
2007	85 (4.07)	.44	1049	6.96	.08
2008	72 (3.45)	–.15	1121	7.02	.07
2009	88 (4.21)	.22	1209	7.10	.08
2010	96 (4.6)	.09	1305	7.17	.08
2011	90 (4.31)	–.06	1395	7.24	.07
2012	93 (4.45)	.03	1488	7.31	.06
2013	97 (4.64)	.04	1585	7.37	.06
2014	113 (5.41)	.16	1698	7.44	.07
2015	133 (6.37)	.18	1831	7.51	.08
2016	135 (6.46)	.02	1966	7.58	.07
2017	123 (5.89)	–.09	2089	7.64	.06

Note. AGR: annual growth rate and RGR: relative growth rate.

documents received a total of 3,358 citations, a mean of 10.17 ($SD = 15.04$) citations per document, and median (Q1–Q3) of 5 (2–12).

The top 10 cited articles and authors in the retrieved literature are shown in Table 5a and Table 5b respectively. The publication that received the highest amount of citations was “Perception of Temporal Patterns” by Povel and Essens (1985), with a total of 364 citations and an average of 11.03 citations per year. The author with the highest number of citations was John Sloboda, who received a total of 1,070 citations.

Table 4. Summary of the citation analysis.

Year	Frequency (%)	TC	Mean (SD)	Median (Q1–Q3)
1973	17 (.81)	86	5.06 (9.17)	1 (0–5)
1974	12 (.57)	118	9.83 (15.44)	2.5 (0.75–10.5)
1975	9 (.43)	25	2.78 (2.86)	2 (0–6)
1976	10 (.48)	60	6 (9.55)	2.5 (0.25–5.75)
1977	11 (.53)	33	3 (2.83)	2 (1–5)
1978	12 (.57)	75	6.25 (8.11)	1.5 (0.75–11.75)
1979	13 (.62)	100	7.69 (12.19)	1 (0–8)
1980	11 (.53)	96	8.73 (10.17)	4 (0–14)
1981	14 (.67)	178	12.71 (13.77)	9 (2–19)
1982	11 (.53)	101	9.18 (10.01)	4 (1.5–15.5)
1983	19 (.91)	272	14.32 (18.3)	9 (1.5–16.5)
1984	38 (1.82)	901	23.71 (27.95)	13 (2.25–33.25)
1985	36 (1.72)	1165	32.36 (65.77)	10.5 (2.75–26.5)
1986	25 (1.2)	507	20.28 (18.12)	18 (8–30)
1987	31 (1.48)	718	23.16 (40.68)	8 (4–19.5)
1988	31 (1.48)	663	21.39 (24.21)	13 (5–29.5)
1989	30 (1.44)	922	22.49 (28.7)	13 (1–31)
1990	35 (1.68)	1045	29.86 (28.41)	16 (11–38.5)
1991	38 (1.82)	1293	34.03 (57.73)	16.5 (9.25–31)
1992	34 (1.63)	628	18.47 (17.58)	16 (4.25–26.75)
1993	28 (1.34)	778	27.79 (38.07)	16 (6–37)
1994	40 (1.91)	875	21.88 (36.17)	14 (6.5–28.25)
1995	34 (1.63)	816	24 (43.25)	14 (3–25.75)
1996	26 (1.24)	942	36.23 (43.43)	21.5 (9.5–50.75)
1997	28 (1.34)	452	16.14 (18.31)	8.5 (3–23.5)
1998	43 (2.06)	808	18.79 (25.39)	11 (2–24.5)
1999	46 (2.2)	1047	22.76 (42.45)	10 (4–22.25)
2000	42 (2.01)	886	21.1 (21.75)	13.5 (7–24.75)
2001	32 (1.53)	996	31.13 (32.66)	24.5 (12.25–38.75)
2002	22 (1.05)	430	19.55 (16.69)	17 (7.25–24.75)
2003	38 (1.82)	812	21.37 (27.93)	7.5 (3–33.25)
2004	32 (1.53)	492	15.38 (30.82)	4 (0.75–14)
2005	57 (2.73)	1308	22.95 (23.87)	17 (7–28)
2006	59 (2.82)	1586	27.34 (29.31)	19 (7.25–27)
2007	85 (4.07)	1978	23.27 (30.1)	12 (5–25)
2008	72 (3.45)	1587	22.04 (21.81)	14 (7.75–27.5)
2009	88 (4.21)	1594	18.11 (22.85)	10.5 (4–23)
2010	96 (4.6)	1443	15.03 (19.11)	8 (3–20)
2011	90 (4.31)	1524	16.93 (21.48)	12 (4.25–17.75)
2012	93 (4.45)	1068	11.48 (10.75)	9 (4–17)
2013	97 (4.64)	888	9.15 (10.45)	6 (3–11)
2014	113 (5.41)	526	4.65 (4.19)	4 (2–4)
2015	133 (6.37)	384	2.89 (3.48)	2 (2–4)
2016	135 (6.46)	173	1.28 (1.62)	1 (0–2)
2017	123 (5.89)	44	.37 (.81)	0 (0–1)

Note. TC: total citations.

Authorship analysis: Productivity, dominance, collaboration, and Lotka’s law

A total of 2,632 authors were covered in the retrieved literature, with a mean of 1.26 authors per article and a mean of .79 articles per author. The mean number of co-authors per article was 2.08. Table 6 shows the average authors per document, author productivity, and collaboration index (CI). The mean number of authors per document increased significantly over time, from a mean of 1.2 in the first period of 10 years (1973–1982) to a mean of 2.48 in the

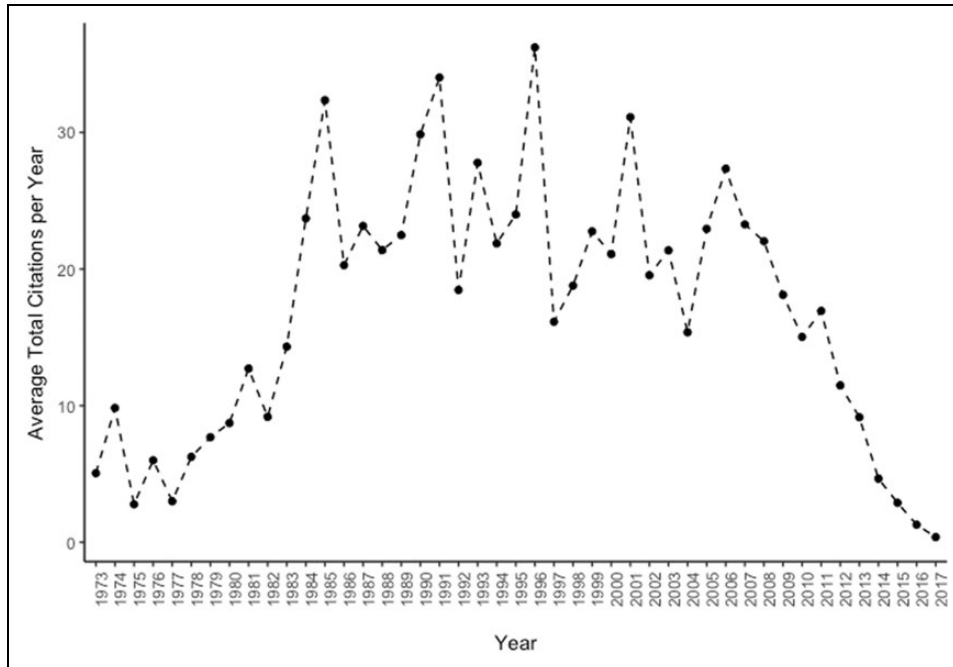


Figure 2. Average total citations per year over time.

Table 5. (a) Top 10 cited articles in the retrieved literature. (b) Top 10 cited authors in the retrieved literature.

Article title	Authors	Journal (year)	TC	TC per year
(a) Top 10 cited articles in the retrieved literature				
Perception of temporal patterns	Povel, D. J., Essens, P.	MP (1985)	364	11.03
Music structure and emotional response: Some empirical findings	Sloboda, J. A.	PoM (1991)	333	12.33
A cross-cultural investigation of the perception of emotion in music: Psychophysical and cultural cues	Balkwill, L. L., Thompson, W. F.	MP (1999)	277	14.58
The emotional sources of “chills” induced by music	Panksepp, J.	MP (1995)	257	11.17
A perceptual model of pulse salience and metrical accent in musical rhythms	Parncutt, R.	MP (1994)	236	9.83
Emotional expression in music performance: Between the performer’s intention and the listener’s experience	Gabrielsson, A., Juslin, P. N.	PoM (1996)	204	9.27
The role of music in adolescents’ mood regulation	Saarikallio, S., Erkkilä, J.	PoM (2007)	203	18.45
Visual perception of performance manner in the movements of solo musicians	Davidson, J. W.	PoM (1993)	197	7.88
Music cognition and perceptual facilitation: A connectionist framework	Bharucha, J. J.	MP (1987)	197	6.35
A model of expressive timing in tonal music	Todd, N.	MP (1985)	193	5.85
(b) Top 10 cited authors in the retrieved literature				
Authors	TC	TP	h-index	
Sloboda, J. A.	1,070	21	14	
Thompson, W. F.	736	16	14	
Krumhansl, C. L.	725	15	13	
Davidson, J. W.	699	20	14	
Eerola, T.	608	17	12	
Hargreaves, D. J.	522	16	12	
Repp, B. H.	515	27	15	
North, A. C.	460	19	11	
Schubert, E.	459	22	11	
Trehub, S. E.	437	17	10	

Note. PoM: *Psychology of Music*; MP: *Music Perception*; MS: *Musicae Scientiae*. TC: total citations; TP: total publications.

Table 6. Average authors per document, author productivity, and collaboration index.

Year	Frequency (%)	TA	Average TA per document	Number of single-authored publications (%)	Number of multi-authored publications (%)	Average TA in multi-authored publications	CI
1973	17 (.81)	17	1.00	17 (100)	0 (0)	0	.00
1974	12 (.57)	13	1.08	11 (91.67)	1 (8.33)	2	2.00
1975	9 (.43)	10	1.11	8 (88.89)	1 (11.11)	2	2.00
1976	10 (.48)	11	1.10	9 (90)	1 (10)	2	2.00
1977	11 (.53)	15	1.36	7 (63.64)	4 (36.36)	8	2.00
1978	12 (.57)	19	1.58	6 (50)	6 (50)	13	2.17
1979	13 (.62)	14	1.08	12 (92.31)	1 (7.69)	2	2.00
1980	11 (.53)	12	1.09	10 (90.91)	1 (9.09)	2	2.00
1981	14 (.67)	18	1.29	11 (78.57)	3 (21.43)	7	2.33
1982	11 (.53)	15	1.36	7 (63.64)	4 (36.36)	8	2.00
1983	19 (.91)	24	1.26	14 (73.68)	5 (26.32)	10	2.00
1984	38 (1.82)	52	1.37	27 (71.05)	11 (28.95)	25	2.27
1985	36 (1.72)	49	1.36	26 (72.22)	10 (27.78)	23	2.30
1986	25 (1.2)	37	1.48	17 (68)	8 (32)	20	2.50
1987	31 (1.48)	48	1.55	21 (67.74)	10 (32.26)	27	2.70
1988	31 (1.48)	55	1.77	16 (51.61)	15 (48.39)	39	2.60
1989	30 (1.44)	50	1.67	17 (56.67)	13 (43.33)	33	2.54
1990	35 (1.68)	62	1.77	18 (51.43)	17 (48.57)	44	2.59
1991	38 (1.82)	72	1.89	18 (47.37)	20 (52.63)	54	2.70
1992	34 (1.63)	59	1.74	16 (47.06)	18 (52.94)	43	2.39
1993	28 (1.34)	54	1.93	15 (53.57)	13 (46.43)	39	3.00
1994	40 (1.91)	61	1.53	25 (62.5)	15 (37.5)	36	2.40
1995	34 (1.63)	47	1.38	24 (70.59)	10 (29.41)	23	2.30
1996	26 (1.24)	48	1.85	15 (57.69)	11 (42.31)	33	3.00
1997	28 (1.34)	42	1.50	17 (60.71)	11 (39.29)	25	2.27
1998	43 (2.06)	89	2.07	17 (39.53)	26 (60.47)	72	2.77
1999	46 (2.2)	76	1.65	27 (58.7)	19 (41.3)	49	2.58
2000	42 (2.01)	68	1.62	20 (47.62)	22 (52.38)	48	2.18
2001	32 (1.53)	57	1.78	15 (46.88)	17 (53.13)	42	2.47
2002	22 (1.05)	40	1.82	11 (50)	11 (50)	29	2.64
2003	38 (1.82)	76	2.00	13 (34.21)	25 (65.79)	63	2.52
2004	32 (1.53)	62	1.94	17 (53.13)	15 (46.88)	45	3.00
2005	57 (2.73)	110	1.93	22 (38.6)	35 (61.4)	88	2.51
2006	59 (2.82)	118	2.00	22 (37.29)	37 (62.71)	96	2.59
2007	85 (4.07)	165	1.94	34 (40)	51 (60)	131	2.57
2008	72 (3.45)	191	2.65	15 (20.83)	57 (79.17)	176	3.09
2009	88 (4.21)	216	2.45	24 (27.27)	64 (72.73)	192	3.00
2010	96 (4.6)	198	2.06	40 (41.67)	56 (58.33)	158	2.82
2011	90 (4.31)	208	2.31	22 (24.44)	68 (75.56)	186	2.74
2012	93 (4.45)	253	2.72	18 (19.35)	75 (80.65)	235	3.13
2013	97 (4.64)	242	2.49	23 (23.71)	75 (77.32)	219	2.92
2014	113 (5.41)	275	2.43	25 (22.12)	88 (77.88)	250	2.84
2015	133 (6.37)	331	2.49	31 (23.31)	102 (76.69)	300	2.94
2016	135 (6.46)	357	2.64	24 (17.78)	111 (82.22)	333	3.00
2017	123 (5.89)	319	2.59	24 (19.51)	99 (80.49)	295	2.98

Note. Percentages in brackets. TA: total number of authors and CI: collaboration index.

last period of 10 years (2008–2017), $F(1,43) = 221.19$, $p < .001$, $R^2 = .837$. The collaboration index (CI) for multi-authored papers (CI = number of authors in multi-authored publications/number of multi-authored papers) increased significantly over time from 2.00 in 1974 (the first year with a multi-authored paper) to 2.98 in 2017, $F(1,43) = 78.91$, $p < .001$, $R^2 = .653$.

Figure 3 shows the number of single-authored and multi-authored publications over time. While a total of 828 documents (39.67%) were single-authored publications, a total of 1,262 publications (60.41%) were multi-authored. Both the number of single-authored papers, $F(1,43) = 27.92$, $p < .001$, $R^2 = .394$, and multi-authored papers, $F(1,43) = 123.97$, $p < .001$, $R^2 = .742$, increased

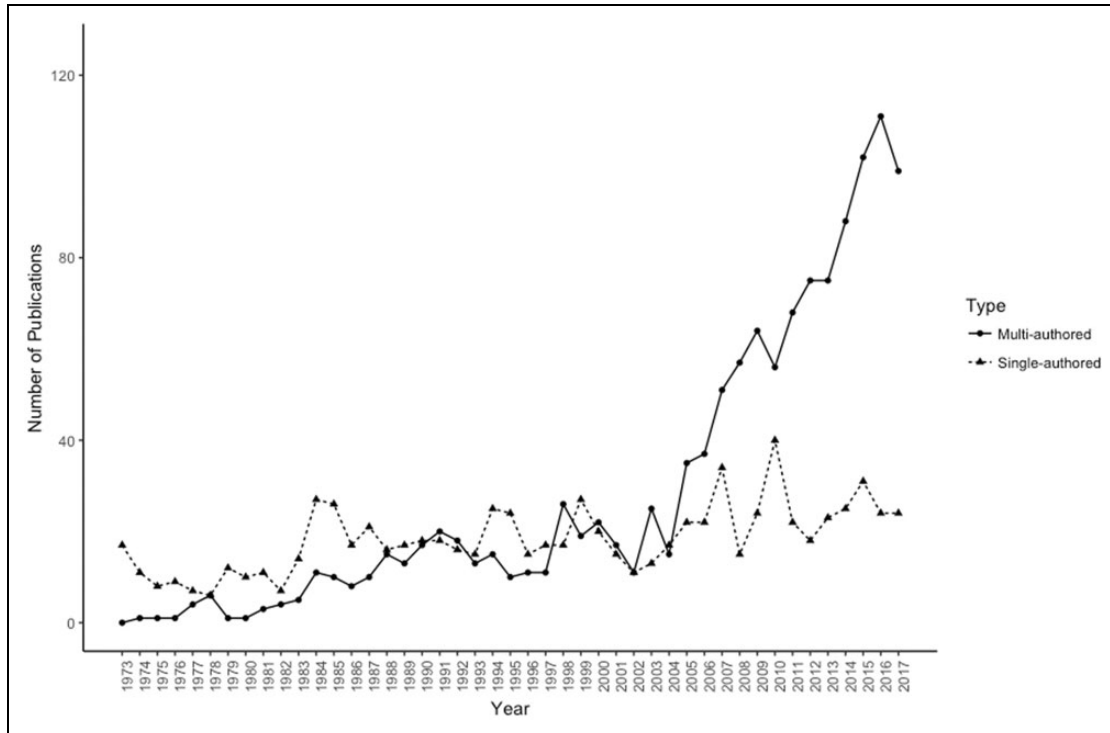


Figure 3. Number of single-authored and multi-authored publications over time.

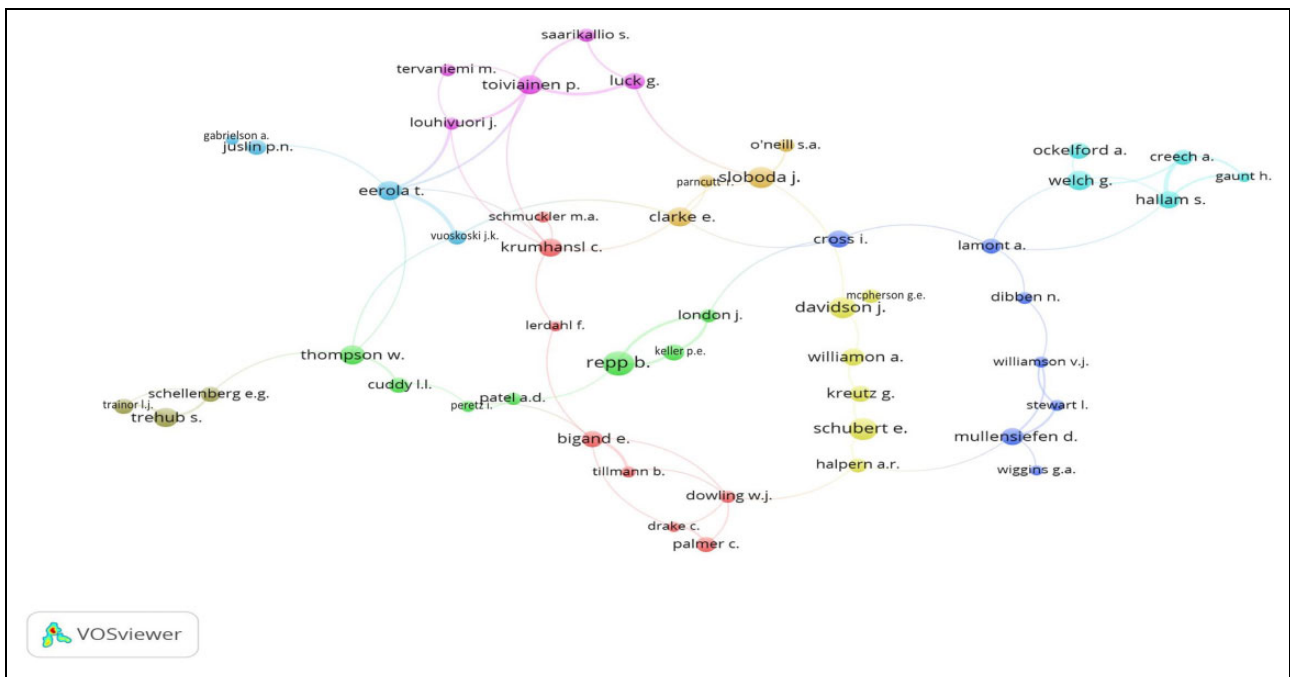


Figure 4. Network visualization map of author collaborations. Note: The width of the line shows the strength of the collaboration. The size of the circle indicates the total number of publications per author. The color of the circle indicates the cluster to which the author belongs.

significantly over time, although this increase had a larger magnitude in publications with multiple authors.

Figure 4 shows a network visualization map of author collaborations. The relatedness of authors is determined

based on their number of co-authored publications. Authors with a minimum of five co-authorship publications and a minimum of 100 total citations are visualized, resulting in a total of 49 authors.

Table 7. Authors with a minimum dominance factor of $> .1$.

Authors	Dominance factor	Multi-authored publications	First author publications
Eerola, T.	.47	17	8
Thompson, W. F.	.37	16	6
North, A. C.	.31	19	6
Repp, B. H.	.30	27	8
Clarke, E. F.	.29	17	5
Müllensiefen, D.	.28	14	4
Kopiez, R.	.27	22	6
Krumhansl, C. L.	.27	15	4
Macdonald, R. A. R.	.25	20	5
Hargreaves, D. J.	.25	16	4
Welch, G. F.	.25	16	4
Sloboda, J. A.	.24	21	5
Hallam, S.	.23	13	3
Davidson, J. W.	.20	20	4
Huron, D.	.14	21	3
Williamon, A.	.14	14	3
Toiviainen, P.	.12	17	2

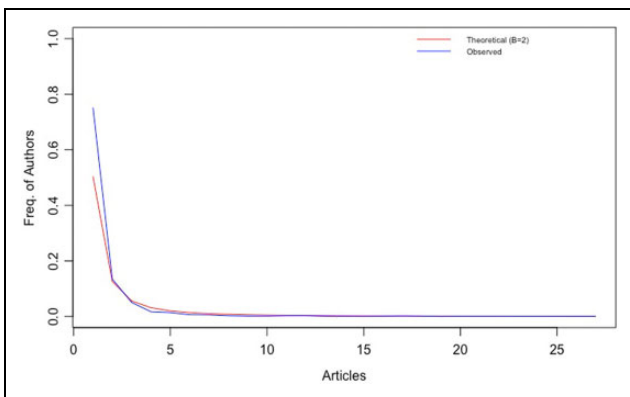
**Figure 5.** Lotka's law coefficient for scientific productivity (theoretical and estimated distributions).

Table 7 shows the authors with a minimum dominance factor of $> .1$. The dominance factor was proposed by Kumar and Kumar (2008), indicating a ratio of the fraction of multi-authored publications in which an author appears as first author (dominance factor 1 means that an author is the first author in all of his or her multi-authored papers). The author with the highest dominance factor (.47) was Tuomas Eerola, being the first author in 8 publications out of 17 multi-authored publications.

Figure 5 depicts Lotka's law coefficient for scientific productivity (Lotka, 1926), indicating the theoretical distribution (red) and the estimated distribution based on the retrieved literature (blue). Lotka's law describes the frequency of publication by authors in any given field. It assumes an inverse square law in which the number of authors making a certain number of contributions is a fixed ratio to the number of authors publishing a single article, implying that the theoretical Beta coefficient of Lotka's law nearly always equals 2. Using the function *lotka* from

the R package bibliometrix (Aria & Cuccurullo, 2017), we estimated the Beta coefficient of the retrieved literature, which was 2.3 and had a goodness of fit equal to .94. A Kolmogorov–Smirnov two-sample test indicated that there were no significant differences between the observed and the theoretical Lotka distribution, $p = .22$.

Country analysis: Productivity, collaborations, and geographical distribution

The number of countries contributing to the retrieved literature was 49. Table 8 displays the countries with a minimum production of five publications, including their frequency, total number of citations, and the number of single-country publications as well as multiple-country publications. The U.S. and the UK had the highest total citations, with 8,669 (25.67%) and 5,954 (17.63%) and a mean of 17.99 and 18.04 citations per publication, respectively. Nevertheless, this analysis did not take the population of each country into account. Thus, we ran a second analysis considering the average population for each country. The average population from 1973 to 2017 per country was calculated using the World Bank population data (<https://data.worldbank.org>). The total number of publications was divided by the average population for each country (rounded to the nearest million) to find the total number of publications per million people. Once population was accounted for, Finland and Australia had the highest total publications, with 15 and 8.1 publications per million, respectively.

Figure 6 shows two geographical distributions of publications, a version without correcting for country population (map on the top) and a version correcting for country population (map on the bottom). The maps were created using the R package *rworldmap*. The map on the top is color-coded using six categories (1 = 0–100, 2 = 101–200, 3 = 201–300, 4 = 301–400, 5 = 401–500, and 6 = 501–600 publications), whereas the map on the bottom used eight categories (1 = 0–2, 2 = 3–4, 3 = 5–6, 4 = 7–8, 5 = 9–10, 6 = 11–12, 7 = 13–14, and 8 = 15–16 publications per million). In the two maps, countries colored in dark blue indicate the highest number of publications, and light yellow colored countries the lowest. Countries with no color indicate that there was no retrieved data from these areas.

Figure 7 depicts a network visualization map of international collaborations. The relatedness of countries is determined based on their number of co-authored publications. Countries with a minimum of 10 international co-authorship publications and a minimum of 100 total citations are visualized. As a result, 19 countries are visualized, clustering in four groups.

Conceptual language

Figure 8 shows an overlay visualization map of author keywords occurrences (i.e., keywords listed by the authors on each publication). Only keywords that occurred a minimum

Table 8. Countries with a minimum productivity of five publications (country of corresponding author).

Country	TP (% , N = 2,089)	SCP (% , N = TP country)	SMP (% , N = TP)	TC (% , N = 33,771)	Average TC per publication
U.S.	482 (23.07)	447 (92.74)	35 (7.26)	8,669 (25.67)	17.99
UK	330 (15.8)	279 (84.55)	51 (15.45)	5,954 (17.63)	18.04
Canada	130 (6.22)	113 (86.92)	17 (13.08)	2,775 (8.22)	21.35
Australia	118 (5.65)	99 (83.9)	19 (16.1)	1,622 (4.8)	13.75
Germany	102 (4.88)	87 (85.29)	15 (14.71)	1,439 (4.26)	14.11
Finland	62 (2.97)	47 (75.81)	15 (24.19)	1,250 (3.7)	20.16
France	56 (2.68)	46 (82.14)	10 (17.86)	704 (2.08)	12.57
Netherlands	47 (2.25)	33 (70.21)	14 (29.79)	1,103 (3.27)	23.47
Japan	41 (1.96)	36 (87.8)	5 (12.2)	517 (1.53)	12.61
Israel	33 (1.58)	31 (93.94)	2 (6.06)	452 (1.34)	13.7
Sweden	31 (1.48)	29 (93.55)	2 (6.45)	775 (2.29)	25
Belgium	29 (1.39)	25 (86.21)	4 (13.79)	485 (1.44)	16.72
Austria	22 (1.05)	13 (59.09)	9 (40.91)	329 (.97)	14.95
Italy	21 (1.01)	17 (80.95)	4 (19.05)	127 (.38)	6.05
Spain	15 (.72)	13 (86.67)	2 (13.33)	69 (.2)	4.6
Georgia	13 (.62)	13 (100)	0 (0)	73 (.22)	5.62
Poland	9 (.43)	9 (100)	0 (0)	115 (.34)	12.78
Switzerland	9 (.43)	5 (55.56)	4 (44.44)	84 (.25)	9.33
Norway	8 (.38)	7 (87.5)	1 (12.5)	65 (.19)	8.12
Greece	7 (.34)	4 (57.14)	3 (42.86)	57 (.17)	8.14
New Zealand	7 (.34)	6 (85.71)	1 (14.29)	67 (.2)	9.57
Portugal	7 (.34)	4 (57.14)	3 (42.86)	34 (.1)	4.86
Turkey	7 (.34)	5 (71.43)	2 (28.57)	54 (.16)	7.71
Hong Kong	6 (.29)	5 (83.33)	1 (16.67)	92 (.27)	15.33
Ireland	6 (.29)	6 (100)	0 (0)	72 (.21)	12
South Africa	6 (.29)	6 (100)	0 (0)	21 (.06)	3.5
Wales	6 (.29)	6 (100)	0 (0)	11 (.03)	1.83
Estonia	5 (.24)	5 (100)	0 (0)	29 (.09)	5.59

Note. TP: total publications, TC: total citations, SCP: single-country publication, MCP: multiple-country publication.

of 10 times were included, resulting in a total of 75 keywords. Note, however, that Scopus only provides author keywords data from 2005 onwards. Thus, the overlay map only displays keywords from 2005 to 2017. Overlay maps are similar to network maps but they are colored based on a given score. The scores used in Figure 8 are based on the average publication year of each keyword. Dark blue represents the oldest average year of publications and red the most recent. The interpretation of the maps is the same as in the network visualization maps.

Discussion

This study aimed to analyze, through visualization and bibliometric techniques, all published literature from *Psychology of Music*, *Music Perception*, and *Musicae Scientiae*. Using all available literature in Scopus, a total of 2,089 publications constituted the retrieved literature, covering a time span of 44 years (1973–2017). Overall, there is a clear increase in music psychology research (i.e., number of publications, authors, and collaborations), with a total growth rate of 11%. The retrieved documents received a total of 33,771 citations ($M = 16.17$, $SD = 26.93$), with a median (Q1–Q3) of 7 (2–20). A total of 2,632 authors were covered in the retrieved literature,

with a mean of 1.26 authors per article and a mean of .79 articles per author. While a total of 828 documents (39.67%) were single-authored publications, a total of 1,262 publications (60.41%) were multi-authored. Both the number of single-authored papers and multi-authored papers increased significantly over time. However, the magnitude of this increase was higher in the publications with multiple authors. The collaboration index (CI) for multi-authored papers (i.e., CI = number of authors in multi-authored publications/number of multi-authored papers) also increased significantly over time, from 2.00 in 1974 (the first year with a multi-authored paper) to 2.98 in 2017. Moreover, the retrieved literature covered a total of 49 countries. The U.S. and the UK were the most productive countries, defined as having the highest number of publications (U.S. = 23% and UK = 16%) and citations (U.S. = 26% and UK = 18%), but when corrected for population Finland and Australia had the highest total number of publications per million (Finland = 15 per million and Australia = 8.1 per million). Finally, the keywords “music” and “emotion” had the highest number of co-occurrences as well as connections with other keywords.

The results of this study present objective and measurable patterns and trajectories seen across the development

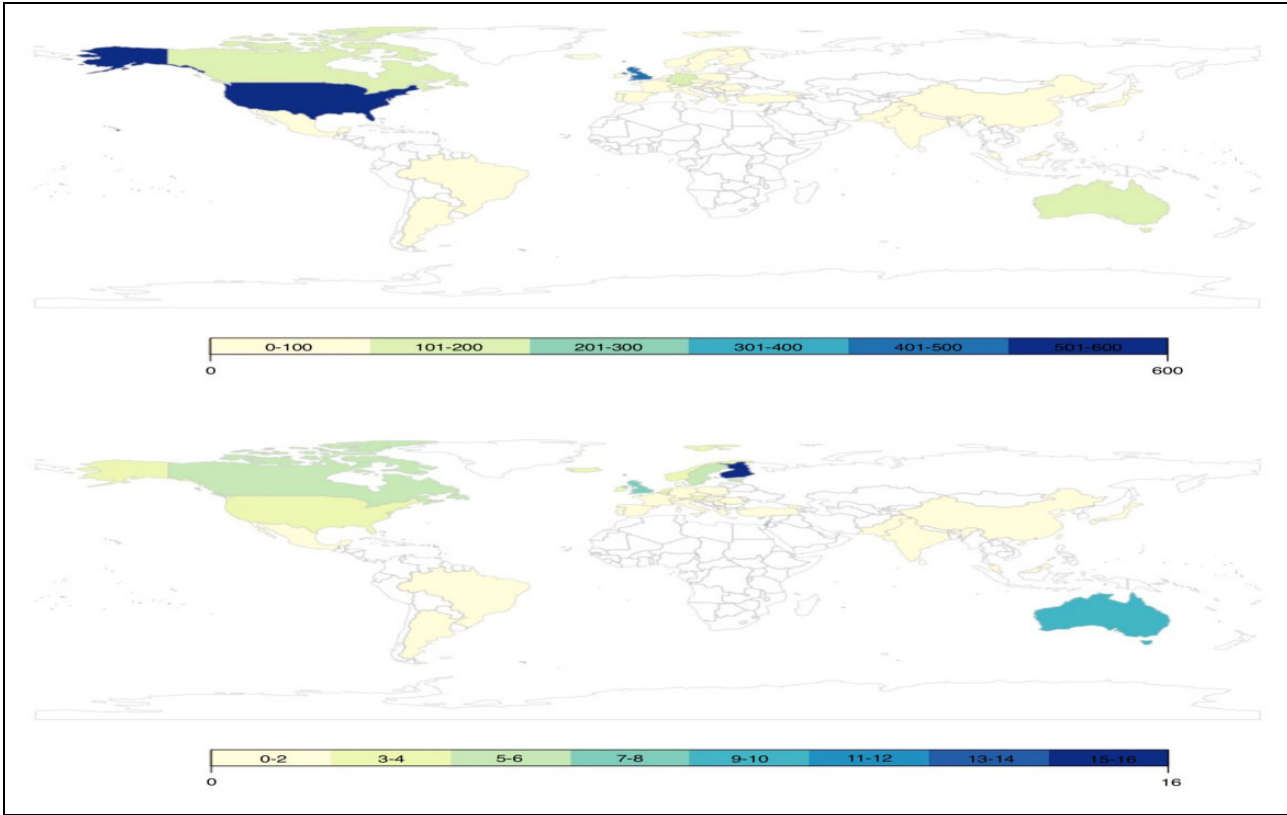


Figure 6. Geographical distribution of publications without correcting for country population (top) and with the correction (bottom). Note: Countries colored dark blue had the highest productivity and countries colored light yellow had the lowest. Countries with no color indicate that there was no retrieved data from these areas.

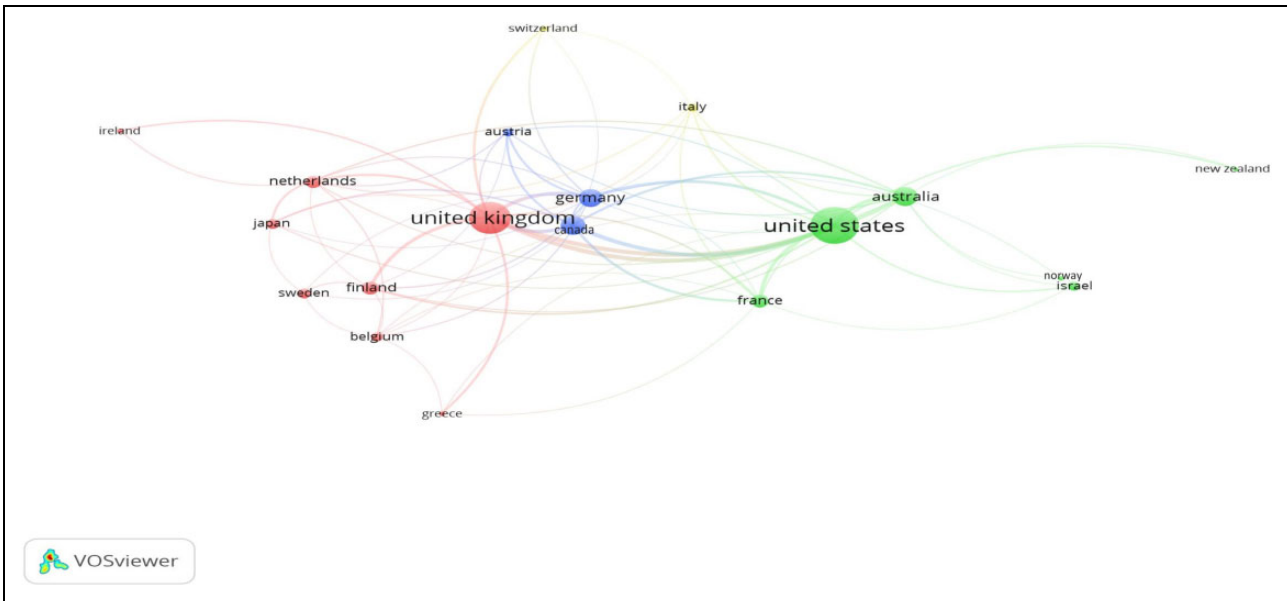


Figure 7. Network visualization map of international collaborations. Note: The width of the line shows the strength of the collaboration. The size of the circle indicates the total number of publications per country. The color of the circle indicates the cluster to which the country belongs.

of music psychology research included within these three journals. We hope these spark discussion and questions as to why these patterns might exist, what gaps they leave, and

how they fit within a wider context. Compared to the above summary of the results, below we discuss these questions for each of our main findings.

of the benefits that the arts can bring to health and well-being. This UK group uses the research findings from music psychology, and other related arts disciplines, to help inform policies. Future research could be done to investigate the subsequent effects of increases in publications on the number of popular science publications and on governmental policies. Understanding this could give better insight into the impact of music psychology research outside an academic audience.

Citation analysis

The retrieved documents received a total of 33,771 citations, with a mean of 16.17 ($SD = 26.93$) citations per document. This is relatively small compared to other related disciplines such as neuroscience, with 187 average citations per article, experimental psychology with 67, and clinical psychology with 68 (Patience, Patience, Blais, & Bertrand, 2017). However, compared to music research publications, which have an average of about seven citations per article, it is relatively higher (Patience et al., 2017).

Across the entire time period, the average number of citations did not increase significantly. However, we identified a significant inverted-U-shaped relationship between year of publication and average number of citations, with its highest peak in 2007, which received 1,978 citations. It is likely that the decrease in the average total citations observed in the last decade is due to the following two factors: an increase in the total number of publications emerging each year and in particular in this last decade; and a natural gap between the year of publication and year of first citation. Hancock and Price (2016) provided some evidence of this gap by examining the first citation speed for articles in *Psychology of Music* from 1973 and 2012. The authors found that the probability of an article receiving a first citation was .25 after 2 years, .50 after 4 years, and .75 after 7 years (Hancock & Price, 2016).

The publication that received the highest amount of citations was “Perception of Temporal Patterns” by Povel and Essens (1985), with a total of 364 citations and an average of 11.03 citations per year. When looking at the top 10 most-cited articles (Table 5a), we see that four out of the ten are about music and emotion and three are about investigating the temporal aspect of music. This may speak to the most-cited areas or sub-disciplines in the field of music psychology within these three journals. The author with the highest number of citations was John Sloboda, who received a total of 1,070 citations. John Sloboda is also known for his research in music and emotion, again emphasizing a key area of music psychology research over the years.

However, note that these results only cover articles published within three music-psychology-specific journals. For instance, we are not capturing articles published in neuroscience or general psychology journals that represent other sub-disciplines within music psychology. It is also important to mention that we only used the citation analysis provided

in Scopus on April 20, 2018. The content of this database is frequently updated, therefore, the numbers reported here will likely change over time. Moreover, there are significant differences between the number of citations indexed in Scopus and other databases, such as Web of Knowledge and Google Scholar (Meho & Yang, 2007). While both Scopus and Web of Knowledge index mostly refereed journal articles, Google Scholar indexes refereed and non-refereed types of documents. In addition, citation counts in different databases rely strongly on the subject matter of the researcher (Meho & Yang, 2007), some subjects being more represented in one database than in another.

Although it was beyond the scope of this study, it would be interesting to carry out an analysis to understand different factors which may predict the number of citations a publication might receive. As predictors, one could use the total number of authors per document, gender of the author, affiliation, country, funding body, research area, and/or journal of publication. For instance, Patience, Patience, Blais, and Bertrand (2017) found that the citation rate correlates positively with the number of funding agencies that finance the research. This is a thought-provoking element we did not account for in the present study. The effect funding has on the dissemination and impact of certain research is known, but not within the field of music psychology specifically.

Authorship analysis

As noted, the magnitude of the increase in publications was higher in the publications with multiple authors and the collaboration index (CI) for multi-authored papers increased significantly over time.

This growth in the total number of authors and collaboration are not just a significant trend in music psychology but are observed in general scientific literature. *The Economist* (2016) found that in 34 million research papers published in peer-reviewed journals and conference proceedings between 1996 and 2015, the average number of authors per paper grew from 3.2 to 4.4. Many factors could be responsible for this growth. One reason could be the fact that research is becoming more multi- and interdisciplinary in general, which is particularly true in the case of music psychology. There is also a growing need for research teams to have different types of expertise and represent a variety of specialist perspectives. This is evident within the field of neuroscience, where the need for inter- and multidisciplinary research has been discussed (e.g., Quagilo et al., 2017, Waldman, 2013). Another reason may be due to authors wanting to “pad their publication lists” and the increasing institutional pressure to “publish or perish” (The Economist, 2016). Multi-authored papers help cut down the workload, resulting in more publications per author per year. Future research could investigate more systematically the reason for this increase and try to understand how this might affect the impact or rigor of published scientific research.

The visualization map also gives a good indication of the spread of collaboration happening both internationally and within specific domains. For example, the blue cluster in the network visualization (Figure 5) includes individuals from a range of sub-disciplines such as everyday uses of music, music perception and music and memory and is mostly comprised of UK researchers. This visualization helps to track how collaborations across different domains and areas may be carried or created by certain dominant individuals within the field.

Finally, when comparing our data set to Lotka's theoretical distribution (Lotka, 1926), we found no significant differences between the observed and the theoretical distributions. Although expected, this is a clear indicator that the literature in music psychology conforms to Lotka's law. That is, the distribution of the number of authors and their scientific productivity (i.e., number of publications) is highly asymmetric: While very few authors publish many articles, the remaining authors publish very few.

Country analysis

When looking specifically at the international collaborations and distributions of publications, we found that out of the total 49 countries contributing to the retrieved literature, the U.S. and the UK were the most productive countries, defined as having the highest number of publications (U.S. = 23% and UK = 16%) and citations (U.S. = 26% and UK = 18%). However, when country population is taken into account, it is Finland (15 per million persons) and Australia (8.1 per million persons) that have the highest publication productivity. By comparing the two maps side by side (Figure 6), the difference can be seen in countries such as the U.S., which, after accounting for population, seems less productive, and countries like Australia and Finland, become the most productive. The large differences in these two scenarios (when not considering country population and when considering it) bring to light the potential of misusing and misinterpreting bibliometric indicators.

The collaboration network map shows this predominance of the UK and the U.S. as well, but also shows how more countries collaborate with the UK, creating more international collaborations than with the U.S. This may have to do with the UK being within the wider EU and thus fostering more collaboration between countries. This prominence of research coming from the U.S. and the UK is not specific to music psychology. However, the full picture of nation productivity in music psychology looks different compared to the general picture. The world's most research-intensive nations, measured by field-weighted citation impact are the UK, U.S., China, Japan, Germany, Italy, Canada and France (Kisjes, 2013). However, in our study the top eight most productive countries were the U.S., UK, Australia, Canada, Germany, Finland, France, and the Netherlands. The productivity of these countries may be related to certain funding opportunities, number of labs and

number of teaching programs based in these countries. Future research could investigate how funding affects the geographical distribution of music psychology. It is important to think about which nation's voices are being heard and which are the loudest within music psychology research. There is a limitation in knowledge if only a few nations are represented. Working towards creating opportunities in other countries for music psychology research and providing places for people to train could help disperse the distribution beyond Europe and the U.S.

Main conceptual language

The keywords that researchers used to describe their articles and how often they co-occur with others indicate the research trends and themes in music psychology. By selecting those keywords that occurred a minimum of 10 times we obtained a total of 75 keywords (Figure 8). The keywords "music" and "emotion" have the highest number of co-occurrences as well as connections with other keywords. This finding is in line with the general interest and significant increase in research on music and emotion (Eerola & Vuoskoski, 2013; Gabrielson & Lindstrom, 2001; Juslin & Laukka, 2003; Västfjäll, 2001). While some keywords connect very well with others (e.g., memory, performance, preference), others are more disconnected (e.g., flow, cross-cultural, musical expertise). It is also interesting to see how a close group of keywords represent research areas. For instance, a clear research area is constituted by "timing", "synchronization", "rhythm", and "meter"; another by "music therapy", "stress", "depression", "individual differences", and "personality". In addition, the overlay map shows how keyword use changes over time. We can see that keywords such as "synchronization" and "timing" both co-occur and are prominently used in the early 2000s, whereas keywords such as "self-regulation", "flow", and "emotion regulation" appear more in recent publications. Overall, this network map allows us to summarize and better understand the complex field of music psychology in a single picture, but the applications of this visualization technique are far-reaching. We encourage researchers to use this tool to define unexplored research areas within music psychology as well as complement their literature reviews. Although this is the first published article that uses VOSviewer (Van Eck & Waltman, 2010) to create visualization network maps within music psychology, the software has been used in more than 500 publications since 2006 (www.vosviewer.com/publications).

Limitations of the study

The present study has two main limitations. First, we only included three journals in our analysis. This choice was based on the journals' content and impact. The aim was to select the most prominent journals that specifically look at music psychology research. Moreover, we needed to use journals indexed in Scopus, as we used this database to

retrieve the literature (e.g., the journals of *Psychomusicology: Music, Mind, and Brain* and *Music & Science* are not indexed in Scopus). This is an important limitation for two reasons. Firstly, these journals are all predominantly English language journals, creating automatically a bias in the types of publications included and researchers represented. Secondly, high-quality research on music psychology is published in a wide range of journals from a wide variety of disciplines, including experimental psychology, social psychology, clinical psychology, computer science, marketing and advertising, personality, and neuroscience. Thus, our study only examines a fraction of the total number of music psychology research publications and our conclusions can then only be drawn from this fraction of literature. It also means that some authors that do not appear as relevant in this dataset might actually be very influential in general.

Furthermore, it is likely that authors working in psychology departments or medical and neuroscience contexts often prefer to publish in mainstream generic journals (e.g., *Journal of Experimental Psychology*, *Journal of Cognitive Neuroscience*) rather than specialist music psychology journals. These generic journals have a higher impact compared to the specialist journals covered in the current study. Therefore, when researchers aim to disseminate their work in the most influential and prestigious journals, they might be encouraged to choose generic over specialist journals. Future research could broaden the scope of the present study by conducting a bibliometric analysis covering music psychology literature in other journals, including generic journals in general psychology, behavioral sciences, medicine, and neuroscience.

The second main limitation relies on the use of Scopus to retrieve the literature, including the citation analysis. This limitation is inherent to any bibliometric study using similar search strategies. Even though Scopus is the largest existing database (Falagas et al., 2008), it is not a complete record of all published literature, due to licensing. For example, articles from *Music Perception* between 2002 and 2004 are missing in Scopus. In addition, when performing databases searches, there is a potential for false positive and false negative results; and the number of citations differ depending on the database (Meho & Yang, 2007). Finally, some authors might have more than one name or different name spelling, which might have caused inaccuracies in the result. Although no ideal solution exists to this problem, we reduced its potential negative impact at the minimal level by deleting the second initials from all authors' names in the retrieved dataset, including only the first surname and the first initial. We hope that the limitations of the current study are justified by the benefits of using large-scale computational bibliometric analysis.

Conclusion

The study reported here begins to investigate the general research trends, reach, and gaps within the published

literature in three prominent music psychology journals. Using bibliometric techniques to visualize and understanding the past and present of research in music psychology leads us to critical observations and conclusions, opening many interesting avenues for future collaborations and research in the field.

More international collaboration outside of Europe and the U.S. should be pursued, allowing for different types of questions, methods and potential findings, steering our field away from WEIRD (Westernized, educated, industrialized, rich, and democratic) populations (Henrich, Heine, & Norenzayan, 2010). Future studies should be done to investigate potential predictors of music psychology research citations. Understanding how the system around music psychology research, its funding schemes, organizations and institutions, and the influence of certain individuals and countries impact the dissemination and academic impact of music psychology research could shed light on how the system is working and potential ways to improve it. Finally, future research should continue investigating the wider impact of music psychology research on the general public and policies. The need for efficient measurements of scientific collaboration and research impact is becoming more important. Using similar large-scale computational analysis allows for these questions to be more objectively addressed.

Music psychology is still a relatively young field. Taking the time to systematically look back and reflect on how the field has progressed, which this study has only just begun to do, helps push the field forward in new and exciting directions. More research, using similar methods, should be done giving insight into the past, present and future of music psychology research.

Acknowledgement

The authors would like to thank Professor Lauren Stewart, Peter Harrison, Tabitha Trahan, Haia Ironside, and Kai Mueller for constructive criticism of the manuscript.

Author contribution

MAT conceived of the idea and the analysis strategy for the study. All other aspects of the research were done collaboratively by MAT and KRMS.


Declaration of conflicting interests

The author declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

Funding

The author disclosed receipt of the following financial support for the research, authorship, and/or publication of this article: This work was supported by a PhD studentship from the "Studienstiftung des Deutschen Volkes" (Bonn, Germany), awarded to Manuel Anglada-Tort.

ORCID iD

Manuel Anglada-Tort  <https://orcid.org/0000-0003-3421-9361>

Peer review

Graham Welch, University College London, Institute of Education.
John Sloboda, Guildhall School of Music and Drama.
One anonymous reviewer.

Supplemental material

Supplemental material for this article is available online.

References

- Aria, M., & Cuccurullo, C. (2017). bibliometrix: An R-tool for comprehensive science mapping analysis. *Journal of Informetrics, 11*, 959–975.
- Bellis, N. De. (2009). *Bibliometrics and citation analysis: From the science citation index to cybermetrics*. Lanham, MD: Scarecrow Press.
- Blázquez-Ruiz, J., Guerrero-Bote, V. P., & Moya-Anegón, F. (2016). New scientometric-based knowledge map of food science research (2003 to 2014). *Comprehensive Reviews in Food Science and Food Safety, 15*, 1040–1055. doi:10.1111/1541-4337.12223
- Blažun, H., Kokol, P., & Vošner, J. (2015). Research literature production on nursing competences from 1981 till 2012: A bibliometric snapshot. *Nurse Education Today, 35*, 673–679. doi:10.1016/j.nedt.2015.01.002
- Chen, S., Arsenault, C., Gingras, Y., & Larivière, V. (2015). Exploring the interdisciplinary evolution of a discipline: The case of Biochemistry and Molecular Biology. *Scientometrics, 102*, 1307–1323. doi:10.1007/s11192-014-1457-6
- Diaz, F., & Silveira, J. S. (2014). Music and affective phenomena: A 20-year content and bibliometric analysis of research in three eminent journals. *Journal of Research in Music Education, 62*, 66–77.
- Eerola, T., & Vuoskoski, J. K. (2013). A review of music and emotion studies: Approaches, emotion models, and stimuli. *Music Perception: An Interdisciplinary Journal, 30*, 307–340.
- Falagas, M. E., Pitsouni, E. I., Malietzis, G. A., & Pappas, G. (2008). Comparison of PubMed, Scopus, web of science, and Google scholar: Strengths and weaknesses. *The FASEB Journal, 22*, 338–342.
- Gabrielsson, A., & Lindström, E. (2001). The influence of musical structure on emotional expression. In P. N. Juslin & J. A. Sloboda (Eds.), *Music and emotion: Theory and research* (pp. 223–248). New York, NY: Oxford University Press.
- Greenberg, D. M. (2017, December 21). *Musical preferences and the brain*. Retrieved from <https://www.psychologytoday.com/us/blog/the-power-music/201712/musical-preferences-and-the-brain>
- Hallam, S., Cross, I., & Thaut, M. (2011). *Oxford handbook of music psychology* (2nd ed.). Oxford, UK: Oxford University Press.
- Hancock, C. B., & Price, H. E. (2016). First citation speed for articles in Psychology of Music. *Psychology of Music, 44*(6), 1454–1470.
- Henrich, J., Heine, S. J., & Norenzayan, A. (2010). The weirdest people in the world? *Behavioral and Brain Sciences, 33*, 61–83.
- Juslin, P. N., & Laukka, P. (2003). Communication of emotions in vocal expression and music performance: Different channels, same code? *Psychological Bulletin, 129*, 770.
- Kisjes, I. (2013, December 10). *Report: How do the large research nations compare?* Retrieved from <https://www.elsevier.com/connect/report-how-do-the-large-research-nations-compare>
- Kumar, S., & Kumar, S. (2008). Collaboration in research productivity in oil seed research institutes of India. In: *Proceedings of Fourth International Conference on Webometrics, Informetrics and Scientometrics*.
- Laengle, S., Merigó, J. M., Miranda, J., Słowiński, R., Bomze, I., Borgonovo, E., ... Teunter, R. (2017). Forty years of the European Journal of Operational Research: A bibliometric overview. *European Journal of Operational Research, 262*, 803–816. doi:10.1016/j.ejor.2017.04.027
- Levitin, D. J. (2006). *This is your brain on music: The science of a human obsession*. New York, NY: Penguin.
- Lotka, A. J. (1926). The frequency distribution of scientific productivity. *Journal of the Washington academy of sciences, 16*(12), 317–323.
- Meho, L. I., & Yang, K. (2007). Impact of data sources on citation counts and rankings of LIS faculty: Web of Science versus Scopus and Google Scholar. *Journal of the American Society for Information Science and Technology, 58*, 2105–2125.
- Mryglod, O., Holovatch, Y., Kenna, R., & Berche, B. (2016). Quantifying the evolution of a scientific topic: Reaction of the academic community to the Chernobyl disaster. *Scientometrics, 106*, 1151–1166. doi:10.1007/s11192-015-1820-2
- Naukkariinen, O., & Bragge, J. (2016). Aesthetics in the age of digital humanities. *Journal of Aesthetics & Culture, 8*, 30072. doi:10.3402/jac.v8.30072
- Patience, G. S., Patience, C. A., Blais, B., & Bertrand, F. (2017). Citation analysis of scientific categories. *Heliyon, 3*, e00300.
- Povel, D.-J., & Essens, P. (1985). Perception of temporal patterns. *Music Perception, 2*, 411–440. doi:10.2307/40285311
- Quaglio, G., Corbetta, M., Karapiperis, T., Amunts, K., Koroshetz, W., Yamamori, T., & Draghia-Akli, R. (2017). Understanding the brain through large, multidisciplinary research initiatives. *The Lancet Neurology, 16*, 183–184.
- Sacks, O. (2007). *Musicophilia: Tales of music and the brain*. Toronto, Canada: Vintage Canada.
- Sloboda, J. A., & Ginsborg, J. (2018). 25 years of ESCOM: Achievements and challenges. *Musicae Scientiae, 22*, 147–160.
- South, A. (2011). rworldmap: A New R package for Mapping Global Data. *R Journal, 3*.
- Sweileh, W. M. (2017). Global research trends of World Health Organization's top eight emerging pathogens. *Globalization and Health, 13*, 9. doi:10.1186/s12992-017-0233-9
- Sweileh, W. M., Al-Jabi, S. W., AbuTaha, A. S., Zyoud, S. H., Anayah, F. M. A., & Sawalha, A. F. (2017). Bibliometric analysis of worldwide scientific literature in mobile - health: 2006-2016. *BMC Medical Informatics and Decision Making, 17*, 1–12. doi:10.1186/s12911-017-0476-7

- Sweileh, W. M., Al-Jabi, S. W., Sawalha, A. F., & Zyoud, S. H. (2016). Bibliometric profile of the global scientific research on autism spectrum disorders. *SpringerPlus*, 5. doi:10.1186/s40064-016-3165-6
- Sweileh, W. M., Shraim, N. Y., Al-Jabi, S. W., Sawalha, A. F., Rahhal, B., Khayat, R. A., & Zyoud, S. H. (2016). Assessing worldwide research activity on probiotics in pediatrics using Scopus database: 1994–2014. *World Allergy Organization Journal*, 9, 25. doi:10.1186/s40413-016-0116-1
- Tan, S. L., Pfordresher, P., & Harré, R. (2017). *Psychology of music: From sound to significance*. London, UK: Routledge.
- Thaut, M. (2011). History and research. In S. Hallam, I. Cross, & M. Thaut (Eds.), *Oxford handbook of music psychology* (2nd ed.). Oxford, UK: Oxford University Press.
- The Economist. (2016, November 25). *Scientific papers get more authors*. Retrieved from <https://www.economist.com/graphic-detail/2016/11/25/scientific-papers-get-more-authors>
- Thompson, W. F. (2009). *Music, thought & feeling: Understanding the psychology of music*. New York, NY: Oxford University Press.
- Tirovolas, A. K., & Levitin, D. J. (2011). Music perception and cognition research from 1983 TO 2010: A categorical and bibliometric analysis of empirical articles in music perception. *Music Perception*, 29, 23–36. doi:10.1525/mp.2011.29.1.23
- van Eck, N. J., & Waltman, L. (2010). Software survey: VOSviewer, a computer program for bibliometric mapping. *Scientometrics*, 84, 523–538. doi:10.1007/s11192-009-0146-3
- Västfjäll, D. (2001). Emotion induction through music: A review of the musical mood induction procedure. *Musicae Scientiae*, 5, 173–211.
- Waldman, D. A. (2013). Interdisciplinary research is the key. *Frontiers in Human Neuroscience*, 7, 562.
- Waltman, L., Van Eck, N. J., & Noyons, E. C. (2010). A unified approach to mapping and clustering of bibliometric networks. *Journal of Informetrics*, 4, 629–635.
- Ware, M., & Mabe, M. (2015). The STM report: *An overview of scientific and scholarly journal publishing*. Retrieved from <http://digitalcommons.unl.edu/scholcom/9>
- Zatorre, R. J., & Salimpoor, V. N. (2013). Why music makes our brain sing. *New York Times*, 9. Retrieved from <https://www.nzcf.org.nz/wp-content/uploads/Why-music-makes-the-brain-sing.pdf>