

# Citation Analysis May Severely Underestimate the Impact of Clinical Research as Compared to Basic Research

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## Abstract

**Background:** Citation analysis has become an important tool for research performance assessment in the medical sciences. However, different areas of medical research may have considerably different citation practices, even within the same medical field. Because of this, it is unclear to what extent citation-based bibliometric indicators allow for valid comparisons between research units active in different areas of medical research.

**Methodology:** A visualization methodology is introduced that reveals differences in citation practices between medical research areas. The methodology extracts terms from the titles and abstracts of a large collection of publications and uses these terms to visualize the structure of a medical field and to indicate how research areas within this field differ from each other in their average citation impact.

**Results:** Visualizations are provided for 32 medical fields, defined based on journal subject categories in the Web of Science database. The analysis focuses on three fields: *Cardiac & cardiovascular systems*, *Clinical neurology*, and *Surgery*. In each of these fields, there turn out to be large differences in citation practices between research areas. Low-impact research areas tend to focus on clinical intervention research, while high-impact research areas are often more oriented on basic and diagnostic research.

**Conclusions:** Popular bibliometric indicators, such as the *h*-index and the impact factor, do not correct for differences in citation practices between medical fields. These indicators therefore cannot be used to make accurate between-field comparisons. More sophisticated bibliometric indicators do correct for field differences but still fail to take into account within-field heterogeneity in citation practices. As a consequence, the citation impact of clinical intervention research may be substantially underestimated in comparison with basic and diagnostic research.

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## Introduction

Citation analysis is widely used in the assessment of research performance in the medical sciences [1]. Especially the *h*-index [2] and the impact factor [3–5] are extremely popular bibliometric indicators. However, the use of these indicators for performance assessment has important limitations. In particular, both the *h*-index and the impact factor fail to take into account the enormous differences in citation practices between fields of science [6]. For instance, the average length of the reference list of a publication is much larger in molecular biology than in mathematics. As a consequence, publications in molecular biology on average are cited much more frequently than publications in mathematics. This difference can be more than an order of magnitude [7].

More sophisticated bibliometric indicators used by professional bibliometric centers perform a normalization to correct for

differences in citation practices between fields of science [8,9]. These field-normalized indicators typically rely on a field classification system in which the boundaries of fields are explicitly defined (e.g., the journal subject categories in the Web of Science database). Unfortunately, however, practical applications of field-normalized indicators often suggest the existence of differences in citation practices not only between but also within fields of science. As shown in this paper, this phenomenon can be observed especially clearly in medical fields, in which the citation impact of clinical intervention research may be substantially underestimated in comparison with basic and diagnostic research. Within-field heterogeneity in citation practices is not corrected for by field-normalized bibliometric indicators and therefore poses a serious threat to the accuracy of these indicators.

This paper presents an empirical analysis of the above problem, with a focus on the medical sciences. An advanced visualization

methodology is used to show how citation practices differ between research areas within a medical field. In particular, substantial differences are revealed between basic and diagnostic research areas on the one hand and clinical intervention research areas on the other hand. Implications of the analysis for the use of bibliometric indicators in the medical sciences are discussed.

## Methodology

The analysis reported in this paper starts from the idea that drawing explicit boundaries between research areas, for instance between basic and clinical areas, is difficult and would require many arbitrary decisions, for instance regarding the treatment of multidisciplinary topics that are in between multiple areas. To avoid the difficulty of drawing explicit boundaries between research areas, the methodology adopted in this paper relies strongly on the use of visualization. The methodology uses so-called term maps [10–12] to visualize scientific fields. A term map is a two-dimensional representation of a field in which strongly related terms are located close to each other and less strongly related terms are located further away from each other. A term map provides an overview of the structure of a field. Different areas in a map correspond with different subfields or research areas. In the term maps presented in this paper, colors are used to indicate differences in citation practices between research areas. For each term in a map, the color of the term is determined by the average citation impact of the publications in which the term occurs. We note that the use of visualization to analyze the structure and development of scientific fields has a long history [13], but visualization approaches have not been used before to study differences in citation practices between research areas. The use of term maps, also referred to as co-word maps, has a 30-year history, with early contributions dating back to the 1980s and the beginning of the 1990s [14–16].

The first methodological step is the definition of scientific fields. This study uses data from the Web of Science (WoS) bibliographic database. This database has a good coverage of the medical literature [17] and is the most popular data source for professional bibliometric analyses. Because of their frequent use in field-normalized bibliometric indicators, the journal subject categories in the WoS database are employed to define fields. There are about 250 subject categories in the WoS database, covering disciplines in the sciences, the social sciences, and the arts and humanities. The analyses reported in this paper are based on all publications in a particular subject category that are classified as *article* or *review* and that were published between 2006 and 2010. For each publication, citations are counted until the end of 2011.

Using natural language processing techniques, the titles and abstracts of the publications in a field are parsed. This yields a list of all noun phrases (i.e., sequences of nouns and adjectives) that occur in these publications. An additional algorithm [10] selects the 2000 noun phrases that can be regarded as the most characteristic terms of the field. This algorithm aims to filter out general noun phrases, like for instance *result*, *study*, *patient*, and *clinical evidence*. Filtering out these noun phrases is crucial. Due to their general meaning, these noun phrases do not relate specifically to one topic, and they therefore tend to distort the structure of a term map. Apart from excluding general noun phrases, noun phrases that occur only in a small number of publications are excluded as well. This is done in order to obtain sufficiently robust results. The minimum number of publications in which a noun phrase must occur depends on the total number of publications in a field. For the three fields discussed in the next section, thresholds between 70 and 135 publications were used.

Given a selection of 2000 terms that together characterize a field, the next step is to determine the number of publications in which each pair of terms co-occurs. Two terms are said to co-occur in a publication if they both occur at least once in the title or abstract of the publication. The larger the number of publications in which two terms co-occur, the stronger the terms are considered to be related to each other. In neuroscience, for instance, *Alzheimer* and *short-term memory* may be expected to co-occur a lot, indicating a strong relation between these two terms. The matrix of term co-occurrence frequencies serves as input for the VOS mapping technique [18]. This technique determines for each term a location in a two-dimensional space. Strongly related terms tend to be located close to each other in the two-dimensional space, while terms that do not have a strong relation are located further away from each other. The VOS mapping technique is closely related to the technique of multidimensional scaling [19], but for the purpose of creating term maps the VOS mapping technique has been shown to yield more satisfactory results, as discussed in detail in Ref. [18]. It is important to note that in the interpretation of a term map only the distances between terms are relevant. A map can be freely rotated, because this does not affect the inter-term distances. This also implies that the horizontal and vertical axes have no special meaning.

In the final step, the color of each term is determined. First, in order to correct for the age of a publication, each publication's number of citations is divided by the average number of citations of all publications that appeared in the same year. This yields a publication's normalized citation score. A score of 1 means that the number of citations of a publication equals the average of all publications that appeared in the same field and in the same year. Next, for each of the 2000 terms, the normalized citation scores of all publications in which the term occurs (in the title or abstract) are averaged. The color of a term is determined based on the resulting average score. Colors range from blue (average score of 0) to green (average score of 1) to red (average score of 2 or higher). Hence, a blue term indicates that the publications in which a term occurs have a low average citation impact, while a red term indicates that the underlying publications have a high average citation impact. The VOSviewer software [20] (freely available at [www.vosviewer.com](http://www.vosviewer.com)) is used to visualize the term maps resulting from the above steps.

## Results

Figures 1, 2, and 3 show the term maps obtained for the WoS fields *Cardiac & cardiovascular systems*, *Clinical neurology*, and *Surgery*. These fields were selected because they match well with our areas of expertise. The maps are based on, respectively, 75,314, 105,405, and 141,155 publications from the period 2006–2010. Only a limited level of detail is offered in Figures 1, 2, and 3. To explore the term maps in full detail, the reader is invited to use the interactive versions of the maps that are available at [www.neesjanvanneck.nl/basic\\_vs\\_clinical/](http://www.neesjanvanneck.nl/basic_vs_clinical/). The webpage also provides maps of 29 other medical fields as well as of all 32 medical fields taken together.

The term maps shown in Figures 1, 2, and 3 all indicate a clear distinction between different research areas. Clinical research areas tends to be located mainly in the left part of a map and basic research areas mainly in the right part, although making a perfect distinction between basic and clinical research areas is definitely not possible. The basic-clinical distinction is best visible in the *Cardiac & cardiovascular systems* and *Clinical neurology* maps (Figures 1 and 2), in which the left part consists of clinical intervention research areas (e.g., cardiac surgery and neurosurgery) while the







available bibliometric indicators, one should be aware of biases caused by differences in citation practices between areas of medical research, especially between basic and clinical areas.

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## Author Contributions

Conceived and designed the experiments: NJE LW. Performed the experiments: NJE LW. Analyzed the data: AFJR RJMK WCP. Contributed reagents/materials/analysis tools: NJE LW. Wrote the paper: NJE LW AFJR RJMK WCP.