

Citation Content Analysis (CCA): A Framework for Syntactic and Semantic Analysis of Citation Content

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This study proposes a new framework for citation content analysis (CCA), for syntactic and semantic analysis of citation content that can be used to better analyze the rich sociocultural context of research behavior. This framework could be considered the next generation of citation analysis. The authors briefly review the history and features of content analysis in traditional social sciences and its previous application in library and information science (LIS). Based on critical discussion of the theoretical necessity of a new method as well as the limits of citation analysis, the nature and purposes of CCA are discussed, and potential procedures to conduct CCA, including principles to identify the reference scope, a two-dimensional (citing and cited) and two-module (syntactic and semantic) codebook, are provided and described. Future work and implications are also suggested.

Introduction

Scholarly impact is usually defined as the extent to which a researcher's work (e.g., an article) has been used by other researchers (Bornmann, Mutz, Neuhaus, & Daniel, 2008). It has been considered an essential factor in estimating the value, credit, and contribution of a certain article, journal, institution, or individual (Brown & Gardner, 1985). In citation analysis, this process is usually equated with citation counts provided by ISI Web of Knowledge, Scopus, Google Scholar, and so forth.

Traditionally, a citation is interpreted as author A being influenced by the work of author B, though without any attempt to specify the strength or direction of that influence. Additionally, it is assumed that each reference has made an equal contribution to the citing article. Therefore, in citation network analysis, citing behavior is usually simplified as a linear relationship—an edge in which node A cites node

B—where nodes can be authors, articles, journals, or institutions (Pinski & Narin, 1976). Based on this simplification, scholarly impact is also reduced to the number of citations. Obviously, the goal of traditional citation analysis is to answer two main questions: whether the two articles are connected through citations and how many citations an article has accrued.

In fact, citing is an inherently complex behavior that is usually triggered by a variety of subjective factors (e.g., authors' intellectual and/or social motivations) and cannot be reduced to a simple linear relationship. Nicolaisen (2007), who has reviewed various theories of citing behavior and citation analysis, concluded that there is a widespread belief that citing behavior can be explained by evolutionary accounts of science and scholarship and that it can be understood in terms of psychology, the normative theory, and social constructivist theory. Small (1976) suggested that citing constitutes an author's interpretation of the cited work, which is a process of meaning creating and symbol making. This process is considered as immaterial and sociologically oriented by Swales (1986). Thus, although by reducing citing to numbers and edges one can obtain a general and broad image of scholarly communication, this simplification risks ignoring the rich sociocultural context of research.

The process of selecting citations is far from random in nature (Cronin, 1981; Small, 2011). It is driven by norms. For example, Cronin (2004, p. 43) speaks of "the normative ghost in the machine" and Cronin (1984) and Small (1976) discuss procedural standards to which scientists typically adhere. We refer to such a set of norms as both individual norms and collective norms. The former explains citing behavior as triggered by individual motivations, whereas the latter suggests that citing can be learned in and shaped by specific groups or domains, that is, how one cites is dependent on the discipline one belongs to (see, e.g., Hellqvist, 2010; Milojević, 2012), or on certain characteristics of authors (their academic age, productivity, and collaborative practices) regardless of their discipline or field (see, e.g., Milojević, 2012).

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With these concerns in mind, we propose citation content analysis (CCA) as a promising addition to traditional citation analysis methods that would enable syntactic and semantic as well as quantitative and qualitative analysis of citation content. Traditional citation analysis is mainly quantitative (e.g., citation counts) and pays no attention to the actual context, whereas classical content analysis (CA) is essentially qualitative (e.g., codebook categories) and rarely applied to citation data. Endeavoring to make the best of both methods and fill the gap, CCA is adapted from CA, but it is not a simple mixture of CA and citation analysis. Instead, CCA is mainly established on two rationales: (a) instead of being weighted equally, citations should be granted different weights under different contexts; (b) qualitative factors (e.g., how one cites) and quantitative measurements (e.g., number of citations) should be incorporated and mutually complementary.

Thus, CCA is conceptualized as an endeavor to describe the contextual relationship between citing and cited works, to indicate the social and intellectual interaction between different authors, to investigate the relative contribution of individual and collective norms to citing behavior, and to understand the nature and function of such behavior. In addition, with advances in computing capability and the growth of digital libraries and repositories, it is possible for CCA, as the next generation of citation analysis, to provide applicable classification schemes, to identify specific citing patterns across different domains so as to facilitate further natural language processing (NLP), and to develop scalable text-mining algorithms to extract associations hidden in large document collections.

In this article, we briefly review the history and features of CA in the traditional social sciences, and its previous application in library and information science (LIS). Based on critical discussion of the limits of citation analysis, we propose that citation content analysis (CCA) should be the next generation of citation analysis that will improve traditional bibliometric research. Endeavoring to establish the theoretical framework of CCA, we discuss the nature and purposes of CCA. Potential procedures to conduct CCA are provided and described. Finally, future work and implications are suggested.

A Theory of Citation: Why Do We Need a New Method?

It has become a convention for scientists and researchers to refer to earlier work (e.g., concepts, theories, methods, equipment, results, conclusions) that relates to, inspires, or is used in their own work (Nicolaisen, 2007). Thus, citations have become intellectual linkages across academic and professional disciplines and can be used to study the nature and the development of different domains. In addition, citations can be studied from various perspectives, from information science (bibliometrics) to linguistics (discourse analysis).

As early as 1986, Swales pointed out one crucial problem in citation studies, that of the existence of two relatively

independent and separate orientations in citation analysis—quantitative description of bibliographical references and qualitative interpretation of the symbolic indication embedded in citations. Information scientists usually focus on citation frequency but rarely take citers' rhetorical and linguistic choices into account, whereas linguists mainly focus on the embedded meanings in sample citations, but fail to investigate the large-scale image by quantitative measures. According to Small and Klavans (2011), such a separation is mainly caused by data availability. Namely, quantitative researchers tend to use a database that represents all sciences, but none of these databases provides full-text data, whereas qualitative researchers tend to use relatively small and homogeneous data, the gathering of which is labor intensive and requires close reading, professional knowledge, and expert judgment (see also McCain & Turner, 1989).

Here (see Table 1) we provide a brief summary of the conceptual origins, basic assumptions, and popular methods of analysis of three main features of citation: (a) numerical, (b) literal, and (c) sociocultural.

The most explicit feature of citation is numerical, which means that citations can be studied quantitatively. This view is closely tied to the idea of science of science, or the idea that we can apply scientific methods to study the phenomenon of science itself. This approach has been widely used in the field of information science and has the counting the number of citations as its basis. The basic assumption here is that research impact is not intangible, but measurable—in a quantitative way.

Citation is also literal. Citation is constructed by words (i.e., language), whereas language is a symbolic representation of concepts and ideas. Words can be used as linguistic cues to suggest a citer's intellectual process, cognitive interaction, attitude, and sentiments. Words can also indicate whether the item in question is new, novel, or important, and thus be used to detect scientific breakthroughs, shifts, or revolutions (Small & Klavans, 2011). The literal feature of citation is neither completely explicit nor implicit. Words can be both explicitly quantified by parsing and counting and carefully examined through a qualitative implicit process to determine their semantic meaning.

The third feature of citation is sociocultural, which is implicit and difficult to obtain either from counting references or from the discourse analysis. The reason is that citation is a complex social system where both individual attributes and social dynamics interact and influence each other. Motivations behind every citation may vary greatly: personalized psychological process (e.g., Nicolaisen, 2007), citers' social environment and cultural background (e.g., Hjørland, 2000, 2002), normative tendency (e.g., Kaplan, 1965) governed by the internal norms in sciences proposed by Merton (1973), or a form of persuasion (e.g., Latour & Woolgar, 1986; White, 2004). There is no an existing method that provides a comprehensive analysis of all three features of citations.

Our framework for syntactic and semantic analysis of citation content draws from existing theories of citing. We

TABLE 1. A theoretical foundation to understand the main features of citation.

Features	Orientation	Conceptual origins	Basic assumptions	Analytical methods
Numerical	Explicit	The measurable nature of science	The number of citations can be considered impact indicators, or signs of breakthroughs.	e.g., Citation counts, citation frequency
Literal	Explicit/implicit	The symbolic nature of words	Words and linguistic choices in the citing text can indicate functions and sentiment of citations.	e.g., Discourse analysis, natural language processing
Sociocultural	Implicit	The individual and social nature of selections	The semantic content of citing contexts can suggest the citing motivations.	e.g., Content analysis, psychological experiments, surveys, interviews

strongly believe that to make further advances in citation analysis, two current orientations, quantitative description of bibliographical references and qualitative interpretation of citation context, need to be combined. Therefore, we suggest a new approach (CCA) that incorporates content analysis (CA) and traditional citation analysis and is capable to comprehensively capture the nature of citation.

Classical Content Analysis (CA): A Flexible Method

As a classical research method, CA has been widely used and well defined in traditional social sciences. It was first used in Europe in the 17th century by the church to systematically examine content of early newspapers, then improved by sociologist Max Weber to study press coverage of political issues in Germany in 1910. From the 1920s to 1950s researchers started to develop the theoretical foundations for CA and applied it to mass communication (e.g., Berelson, 1952). Since the 1960s, CA has been extended and applied to other areas, for example, anthropology, history, library and information science, linguistics, management, political science, psychology, and sociology. In this process, researchers from different domains adapted CA to their unique research questions and goals. CA has become “a broadening of text aspects to include syntactic, syntagmatic, and pragmatic aspects of text, although not always within the same study” (White & March, 2006, p. 23). Not surprisingly, in today’s digital era, CA is usually considered a flexible research method with the potential to incorporate both quantitative and qualitative approaches, conducted both manually and with computer assistance, which can be applied to many questions in different domains.

There are multiple definitions of CA reflecting its historical development and rich variants (e.g., conversational analysis, discourse analysis, ethnographic analysis, functional pragmatics, rhetorical analysis, and narrative semiotics; see Krippendorff, 2004), however, here we summarize it in terms of the following four characteristics: dynamics, resource, structure, and operationalization.

Dynamics: Systematic and Objective

Previous studies define CA as a systematic and objective research method. For example, Bauer (2000) identifies it as

a systematic, replicable technique for coding data found in communication (of any type). Holsti (1969) suggests that CA is any technique for making inferences by objectively and systematically identifying specified characteristics of messages. Here “systematic” indicates at least two senses: the systematic process of sampling of messages, and the systematic structure (e.g., symbols-numbers, words, letters, computer codes, etc.) of sampled messages. “Objective” suggests that the analysis should make replicable, repeatable, and valid inferences from texts (or other meaningful matter) to the contexts of their use (Krippendorff, 2004). In this sense, CA is not a subjective interpretation of others’ works but an incorporation of both quantitative and qualitative methods.

Resource: Message-Based

In traditional social science research, CA is based on textual materials. Stone, Dunphy, Smith, and Oqilvie (1966) propose that the ultimate goal of CA is to identify “specified characteristics within text” (p. 5), and to make specific inferences from text to other states or properties of its source. Therefore, CA was mainly used to systematically classify and count textual (word-based) units. However, in today’s digital era the application of CA has been greatly expanded to diverse resources (e.g., images, videos, hyperlinks, etc.) besides pure texts. For example, Kress and van Leeuwen (1996), as well as Bell (2001), provide a framework of visual content analysis of images. Generally, the rich context and wide application of CA have led to its wide use with all kinds of symbolic data (messages in general).

Structure: Syntax and Semantics

As a method embodying quantitative and qualitative components, CA focuses on both syntactic and semantic structures. The former refers to how symbolic data are organized and presented (e.g., feature/image/word frequencies, linguistic indicators, order of elements), whereas the latter demonstrates what is presented (meaning, both denotation and connotation), for example, themes, valuations and so forth. Syntactic and semantic structures are also called “analytical constructs, or rules of inference” (White & March, 2006, p. 27), which can be both quantified and qualified. It is based on these two analyzable structures that implicit

meaning (i.e., content) that is embedded in the explicit presentation (messages) can be interpreted and understood.

Operationalization: Coding

Constructing a systematic classification of message-based units is crucial for CA, one in which coding plays the central role. Cartwright (1953) even proposes that the terms *content analysis* and *coding* can be used interchangeably to emphasize the objective and systematic description of any symbolic behavior. According to White and March (2006), coding constitutes the body of CA and includes (a) establishing a coding scheme that allows for testing hypothesis, (b) coding data, (c) checking the reliability of the coding, (d) adjusting the coding process if necessary, (e) analyzing coded data, and (f) applying appropriate statistical test(s). Not surprisingly, for a number of reasons the most important step is establishing an appropriate coding scheme: (a) it is the coding scheme that operationalizes and qualifies the intangible concepts and implicit connotations; (b) valid and consistent assessment is achieved by establishing a coding scheme with clearly defined, comprehensive and mutually exclusive categories that represent relevant aspects (i.e., facets) of the data; (c) the reliability of research results and conclusions is highly correlated to the appropriateness of a coding scheme. Namely, the better a coding scheme is, the higher the interrater and intrarater reliability is (i.e., different coders will code the same item the same way or a single coder will code the same item the same way at different points in time (see Krippendorff, 2004). In general, constructing such an appropriate scheme is a complex and mainly qualitative process, which often involves careful, repetitive reading of the original messages, and modifying/remodifying of the proposed scheme.

Content Analysis in Library and Information Science

Traditionally, CA has been used to determine authorship (from identifying personalized linguistic and rhetorical characteristics), examine patterns in documents, infer psychological or emotional states, and product evaluation. In library and information science, CA has been extended to analyze different types of data (e.g., reference interviews,

problem statements in published articles, job advertisements, etc.) in both qualitative and quantitative researches. For example, Pettigrew and McKechnie (2001) used a CA codebook (including three categories: affiliation of first author, primary subject of the article, and type of article) to analyze authors' use of theory in 1,160 articles that appeared in six information science journals from 1993 to 1998. In 2006, White and March (2006) provided a summary of selected examples of CA studies in LIS from 1991 to 2005, including identifying the reasons for selecting initial strategy in web searches (White & Iivonen, 2001), developing a thesaurus of image-text relationships (Marsh & White, 2003), determining the nature of problem statements in LIS articles (Stansbury, 2002), and so forth.

In essence, the popularity of CA in LIS originates from its flexibility and appropriate match with the nature of LIS, which is shown in Table 2.

As we mentioned in earlier section, there is one area of LIS where CA is still not widely used—citation analysis, as it is difficult to apply the qualitative essence of CA (e.g., codebook categories) to citation data. This is despite the fact that the idea of combining bibliometric methods with the full text analysis for the purposes of “context and content analysis of citations” (Cronin, 1984, p. 35) was put forward and experimented with as early as the 1960s (Glenisson, Glänzel, Janssens, & De Moor, 2005). For example, in 1965 Lipetz identified 29 different reasons for citing and grouped them into four clusters. In the 1970s a number of researchers (e.g., Chubin & Moitra, 1975; Frost, 1979; Moravcsik & Murugesan, 1975; Oppenheim & Renn, 1978; Spiegel-Rosing, 1977) followed his ideas and proposed their own schemes to categorize and contextualize citations. Small (1982) and Cronin (1984) provided overviews of citation classification schemes. Some of the previous endeavors also focused on cword analysis (e.g., Callon, Courtial, & Laville, 1991) or word analysis (e.g., Braam, Moed, & Van Raan, 1991) in the context of evaluative bibliometrics to improve efficiency of cocitation clustering. However, these approaches are not actual CA.

In summary, the difficulties of combining CA and citation analysis are of two kinds. First, citing behaviors are usually simplified as a linear one-dimensional relationship whereas CA is a descriptive and multidimensional method. Traditional citation analysis assumes that author A has been influenced by the work of author B, though without any attempt

TABLE 2. Advantages of content analysis (CA) in library and information sciences (LIS).

	LIS features	Advantages of CA
Main data type	Raw data, existing historical data, archival records	Well suited to historical data and archival records
Data amount	Usually large amounts of data	Can deal with large amounts of data
Procedure	Replicable, retrievable, recordable	Offers a set of mature and well-documented procedures
Cost	Inexpensive, requires no contact with people	Inexpensive, requires no contact with people
Boundary	Flexible and interdisciplinary	Highly flexible, can be combined with other research methods (e.g., interviews, observation, statistics)

to specify the strength or nature of that influence. Additionally, it is assumed that each reference has made an equal contribution to the citing article. In contrast, CA endeavors to describe the citing behavior itself, as well as to interpret and understand the underlying motives for the observed pattern. Namely, it seeks to understand what the purposes, functions, attitudes, dispositions, and sentiments behind the citing behavior are and how these patterns are represented in citations to indicate authors' motivations. Second, it is difficult to establish an appropriate coding scheme for citing behaviors. As we discussed above, the most important step in CA is the creation of an appropriate coding scheme, which will establish a set of clearly defined, comprehensive, and mutually exclusive categories. One reason for this difficulty lies in identifying sampling units, data collection units, and units of analysis, which constitute the foundation of generating a coding scheme. According to White and March (2006), sampling units serve to identify the population and establish the basis for sampling. Data collection units are the units for measuring variables. Units of analysis are the basis for reporting analyses. In the context of citing behavior, to determine these units we need to make decisions regarding the following: Should all scholarly work, or works in a given domain/discipline, be identified as sampling units? Should long articles, short articles, journal articles, conference papers, or books, be identified as data collection units for measuring? Should a single sentence, or a cluster of sentences in which a reference is mentioned, be selected as the unit of analysis? Confusion generated by these questions indicates that the research domain (and its accepted writing pattern), the dominant genre(s), and the length/coverage of analytical units can influence the creation of a coding scheme, and can constrain the scope of its application as well. In addition, citing behavior can be triggered by subjective factors. As Small (1976) suggests, citing constitutes an author's interpretation of the cited work, which is a process of meaning creating and symbol making. This process is considered as immaterial and too sociologically orientated by Swales (1986). Thus, it could be difficult to "reinterpret" authors' "interpretations" without deep background knowledge of the field and authors themselves.

Nevertheless, these complexities and challenges are not grounds to avoid CA in citation analysis. Pioneering researchers from the 1960s to 1980s have done inspiring works in this area, providing sound foundation and increasing "our understanding of the relationships which exist between citing and cited documents in the scientific literature" (Cronin, 1984, p. 49). As followers of these pioneers, as well as with the aspiration to further investigate the norms and behaviors surrounding citations, we propose a framework for a new method of citation content analysis (CCA) that would introduce CA to the traditional citation analysis in a way that could revolutionize traditional bibliometric research. In the next sections, we discuss the nature and purposes of CCA, and propose potential procedures to conduct CCA.

Citation Content Analysis (CCA): Nature and Purposes

Although the term *citation content analysis* or similar terms have been mentioned in several previous works (e.g., "content citation analysis," Swales, 1986; "citation content analysis," McCain & Turner, 1989), CCA in this article introduces new implication and significance. Namely, CCA is not merely a text/word-based linguistic or discourse analysis approach. It is an endeavor to investigate all three features of citation: numerical, literal, and sociocultural.

A Discourse Approach for Academic Writing

The main reason why CCA can become an appropriate method to analyze citing behavior comes from the nature of academic writing itself. It has been accepted and validated that CA is the most efficient when applied to semantically rich and logically consistent texts (e.g., Markoff, Shapiro, & Weitman, 1975). Academic writing meets all these requirements because it is formal, official, systematic, and neutral to a great degree. Therefore, CCA is well suited for the analysis of texts of such a unique writing style.

In this sense, CCA can effectively organize, standardize, and categorize both the explicit format and the implicit function of texts, so as to conduct systematic comparisons and reasonable interpretation. The coding procedure in CCA can divide, categorize, and transform "this mass of documentation into an organized data file" (Markoff et al., 1975, p. 3), which is highly detailed and concrete. This process of operationalization can facilitate comprehension of intricate texts and promote communication between different researchers to investigate the same data file so as to shed light on the embedded motivations and connotations behind citations and citing behaviors. As Cronin (1981) points out, although such textual analysis cannot tell us all reasons why an author cites as he does, it may suggest very plausible reasons.

A Symbolic Approach for Conceptualizing Citations

Another reason why CCA can be and should be used to investigate citing behaviors is embedded in the symbolic nature of citations. Gilbert and Woolgar (1974) have distinguished citation from reference. Reference refers to the works mentioned in the reference section or bibliography of a journal article. A reference may be mentioned once or multiply in an article. Each mention is considered a citation. Thus, citations are the contexts in which references are made. According to Small (1978), citations can be considered to be "symbols of concepts or methods"—so citing is a process of creating cognitive links between concepts, procedures, types of data, and documents. This view also echoes Garfield's (1977) notion of cited documents as subject headings in an indexing system, and Gilbert's (1977) idea of citing as an author's device for persuading readers. As Cronin (1981) states, "citations are frozen footprints in

the landscape of scholarly achievement; footprints which bear witness to the passage of ideas” (p. 16). The CCA can be used to operationalize and measure the intangible concepts and connotations, as well as the intellectual process of knowledge transfer and sharing. The proposed coding scheme and the analytical procedures can lead to a clearer and more specific image of interactions, conflicts, dialogs between different authors, documents, ideas, and paradigms, than traditional citation analysis.

A Macro-Economic Approach to Citing Behaviors

In principle, CCA conceptualizes citing behavior as a decision-making process in which citing is a way of information aggregation. From the perspective of macro economic theory, citing behavior is viewed as the author’s effort to effectively accomplish a task and to make accurate predictions (Bacon, Chen, Kash, Parkes, Rao, & Sridharan, 2012), and a process of selection to reduce risks and optimize potential output. For example, what is the possibility of gaining acceptance/acknowledgment if I choose to cite author A? What if I cite B instead? What if I cite both A and B? What if I cite A and/or B in different ways? For individuals (authors), this is a prediction problem (e.g., whether or not I will benefit from citing this one) and a selection problem (e.g., Which work I should cite to facilitate my success?) For collectives (a certain domain/field), this is an interaction problem between agents (personal motivations of members) and the community as a whole (established conventions of the field): What is the dynamic embedded in the “outcomes-based incentive system” (Bacon et al., 2012, p. 701) consisting of separate individuals? Using this approach, Othman and Sandholm (2010) have developed a single, deterministic decision rule: Always select the action with the greatest probability of success, as suggested by Chen and Kash’s (2011) study.

From this perspective, citing is not a random behavior or simply piling of all related works, especially considering the existing enormous literature corpus, the trend of interdisciplinary borrowing, and the regular limits of page numbers for publications. Instead, citing is a rational, selective, and comparative way to make best “economic” benefit. It is one way to decrease probability of failure, but increase probability of success, decrease the risk (e.g., rejections, challenges, etc.) and cost (e.g., time, energy, social, cognitive, etc.), but increase security (e.g. acceptance, acknowledged, etc.) and output.

An Indicative Approach to Citing Motivations

Citing is a complex social and academic phenomenon that can be triggered by various subjective factors and cannot be reduced to linear relationship. Therefore, motivations behind citing behaviors, which are embedded in broad social contexts, cannot be interpreted merely by counting the number of citations. Instead, CCA, with its theoretical and

analytical roots in sociology and linguistics, and a grounding in actual discourse can provide a descriptive approach to indicate the in-depth citation motivations based on a broader context.

Unlike business and marketing activities, the “economic” benefit in this context is social rather than financial or monetary capital. Sociologists (e.g., Coleman, 1988; Portes, 1998; Putnam, 1995, 2000) have discussed the origins, definitions and applications of social capital, regarding it as a collective-based and intangible capital, which is generated by networks of relationships, reciprocity, trust, and social norms. Social capital facilitates both individual and collective action. Generally, social capital refers to the value of, and the economic (not monetary) benefits derived from the network of social relationships.

Based on this understanding, acquiring social capital can become an important motivation for citing and citation selections. Essentially, citing is a process of information aggregating to go beyond the limits of personal cognitive capabilities. Today, scholarly collaboration, scientific writing has become a dynamic process of borrowing, incorporating, creating, and improving. Therefore, borrowing from others and self-creating based on previous works have been two crucial components of any scholarly work, as maintaining intellectual consistence and generating originality are equally important for any scientific researcher. As Chen and Kash (2011) state, “[i]nformation is often possessed by individual agents. Truthfully eliciting such information, resolving conflicting beliefs, and aggregating the dispersed information are key problems for achieving collective intelligence in multi-agent systems” (p. 1). Citing establishes the network of collaboration among different researchers, creates social capital in the forms of shared information, understanding, and knowledge, allows them to widen their horizons of understanding, increase their personal access to information and resources, achieve better outcomes, and in turn enhanced power (e.g., greater impact, higher reputation, and broader acknowledgment).

By means of citing, authors as decision makers both predict and influence their future impact. Similar to those decision makers in economic markets as discussed in Chen, Gao, Goldstein, and Kash’s (2011) work, authors can calculate the effects of each set of possible actions—in this case, a set of all possible works an author can cite. This calculation is based on a process of cognitive evaluation in which authors pose questions such as “Is this an appropriate work for me to cite?” “Can I incorporate this work into my work?” “What kinds of benefits can I get from this work?” “How can I cite it to fulfill different purposes?” After this reviewing, authors as decision makers can select an action to perform—citing A or B, or both A and B.

Potential Procedure for CCA

We have discussed the complexities and challenges of incorporating both the descriptive, essentially qualitative

CA method and the linear, simplified, and mainly quantitative citation analysis method. Here we provide a framework for the potential procedure for CCA, endeavoring to optimize advantages of traditional CA and citation analysis, as well as to decrease their limitations.

1. Identify Reference Scope

As noted, the fundamental challenge to create an appropriate CCA method is to identify sampling units, data collection units and units of analysis, all of which constitute the basis of an applicable coding scheme. In other words, the main question is how to determine the reference scope. We propose three principles:

- a. *Principle of diversity* refers to the selection of heterogeneous sampling units, endeavoring to guarantee the generalization of the coding scheme. For example, one should use resources from different scientific domains (e.g., natural sciences, social sciences, humanities, etc.)
- b. *Principle of consistency* refers to the selection of homogeneous data collection units, endeavoring to maintain the comparability of the coding scheme. For example, one should use the same genre (e.g., all conference papers, journal papers, or books.)
- c. *Principle of flexibility* refers to the flexible scope of units of analysis, depending on syntactic or semantic categories in the coding scheme. One should use either single-sentence level or sentence-cluster level. At the single-sentence level, only the citing sentence that mentions previous work will be coded to identify the syntactic features of the citation (e.g., types of cited documents). At the sentence-cluster level, surrounding context (e.g., 1 to 2 sentences before or/and after the exact citing sentence) will be coded to indicate the semantic features of the citation (e.g., functions of citation). In this process, text-mining algorithms, natural language processing (NLP), and topic modeling techniques can be used to determine and identify the scope of a cluster of sentences that are related to a given target reference.

2. Create the Codebook

The greatest challenge is to create an appropriate and applicable codebook for citations, which should be comprehensive but not too complicated, specific but not too detailed, be broadly applicable but not too general. Traditionally, a major criterion for evaluating social science research is its generalizability—the capability to draw reliable conclusions about the whole population based on results of the sample data, that is, inference from the specific to the general (White & March, 2006). However, when creating an appropriate and applicable coding scheme for CCA, the question becomes how to balance specificity and generalizability.

Table 3 is a summary of main coding schemes in citation analysis. The main problem with the existing coding

schemes is their exhaustivity (i.e., researchers tend to create too many categories [more than 20] trying to capture all the possibilities). Such level of details tends to shrink the application scope of their schemes, and imposes enormous pressure on computer-assisted analysis.

In addition, although many LIS studies have been done to combine traditional bibliometric methods with full-text analysis, and to develop classification schemes for citations, previous coding schemes tend to be subject to the following weaknesses: (a) Although addressing the same problem, the previous coding schemes tended to be disconnected from one another and to present different foci (Cronin, 1984). Some schemes focused on the function and quality of citations (e.g., Moravcsik & Murugesan, 1975; Oppenheim & Renn, 1978; Spiegel-Rosing, 1977), some emphasized the reasons for citing (e.g., Lipetz, 1965), and others paid attention to citation sentiment by identifying cue words (Small, 2011; Teufel, Siddharthan, & Tidhar, 2006). (b) The previous coding schemes were constructed more from the perspective of users' needs and perceptions rather than from those of the citing authors, especially in terms of authors' citing motivations. Thus, the contextual interrelations between the citing and cited works, and the distinctive features of the citing and cited authors are vague. Such an ambiguity poses difficulties for the explicit distinction between the citing and the cited, as well as on conducting an in-depth analysis of their interaction.

With a critical review of existing schemes, we propose a two-dimensional (citing and cited) and two-modules (syntactic and semantic) codebook for CCA. Based on grounded theory (Glaser & Strauss, 1967), the key approach we used to create our codebook was to learn from previous schemes and adapt to new challenges, to support both quantitative and qualitative measures, to distinguish citing-generated and cited-generated elements, to indicate both explicit and implicit principles, and to be inclusive of all the formats of resources in different domains. This codebook is also an attempt to balance specificity and generalizability, as well as to investigate the interaction between individual norms (e.g., personal motivations) and collective norms (e.g., established regulations/conventions in a certain domain) in citing behavior. Table 4 summarizes this proposed codebook. In principle, this codebook is three-way orientated: attributes of the citing articles (e.g., Category G, H, K, L), attributes of the cited articles (e.g., Category A, B), and the attributes of the citing-cited interaction (e.g., Category C, D, E, F, I, J).

Depending on categories of the coding scheme, analysis can be conducted at either the syntactic or semantic level. For the syntactic module, analysis is usually conducted at the single-sentence level, and can support traditional bibliometric research on authorship, frequency, and other quantitative measures.

Most important, Category C (relation to the citing work) can be used to study the latent connection between cited and citing works. This category is constructed based on theories of macroeconomics and social capital, implying the potential social triggers embedded in citing behaviors. It contains

TABLE 3. Summary of main coding schemes in citation analysis.

Principles of coding	Sources	Example codes
Type of motivation	Lipetz (1965)	Group 4: Disposition of the scientific contribution of the cited article to the citing article (1–18 are included in Group 1–3, which are not about type of motivation): 19. Noted only 20. Distinguished 21. Reviewed or compared 22. Applied 23. Improved or modified 24. Replaced 25. Changed the precision (plus or minus) 26. Questioned 27. Affirmed 28. Refuted
Level of importance	Moravcsik & Murugesan (1975)	1. Conceptual or operational 2. Organic or perfunctory 3. Evolutionary or juxtapositional 4. Confirmative/negational.
Type of source	McCain & Turner (1989); Frost (1979)	McCain & Turner (1989): 1. Research reports 2. Review articles Frost (1979): 1. Primary sources 2. Secondary sources
Function of citing	Oppenheim & Renn (1978); Spiegel-Rosing (1977)	1. Methodological function (e.g., providing data, developing methods, etc.) 2. General function (e.g., historical background of a subject domain)
Type of disposition/sentiment	Frost (1979); Teufel et al. (2006); Small (2011)	Frost (1979): approval or disapproval; Teufel et al. (2006): Weakness, contrast, positive, neutral Small (2011): Importance, utility, report, consensus, uncertainty, differentiation, negation
Location of mentioning	Herlach (1978), Voos & Dagaev (1976)	Herlach (1978): Title/introduction, results/discussion, experimental Voos & Dagaev (1976): Introduction, method, discussion, conclusion

three categories: reciprocal, parallel, and hierarchical. Reciprocal generally refers to self-citation, which suggests self-acknowledgment/development. Parallel refers to citing peers, coauthors, or collaborators, with a potential intention that the cited author(s) may cite back or reinforce possible collaborations in the future. For example, a sentence in one of Hjørland's works is "Hjørland's (2009) criticized this approach in information science and began developing an alternative 'domain analysis' (Hjørland & Albrechtsen, (1995)." The two citations used in this sentence can be coded as "C1" (reciprocal) and "C2" (parallel), respectively, because the first citation is a self-citation and the second one cites a collaborative work. Hierarchical refers to citing prestigious authors with high social capital, potentially increasing one's own social or scholarly capital. To operationalize this coding, we propose that network analysis should be used to suggest a certain author's social capital based on betweenness, closeness, and degree centrality. By comparing the citing author's and cited author's social capital, one can decide whether the C2 (parallel) or C3 (hierarchical) category should be assigned.

For a given cited work, identifying its location of mentioning (Category D) and counting its frequency of mentioning (Category E) in the same citing article can suggest its level of significance, as well as the different citing patterns across disciplines. As Voos and Dagaev (1976) report, different disciplines exhibit different citation patterns in terms of the locations of citations. They argue that the contribution of a cited reference can be calculated based on the number of times it is cited and the location of those recitations in the citing article and find that highly cited articles appear most often in the introduction. In a similar vein, Herlach (1978) maintains that if a work has been cited in the introduction or literature review section and is mentioned again in the methodology or discussion sections, it is likely that it makes a greater overall contribution to the citing article than others that have been mentioned only once. Thus, it is possible that a reference that was mentioned more than five times in different sections of an article is more important than a reference that was only mentioned once at the very end.

TABLE 4. Two-dimensional and two-modular codebook for citation content analysis.

		Syntactic (Sy)	
Orientation	Categories	Values	
Cited	A. Type of cited documents	1. Journal article	
		2. Conference paper	
		3. Book/book chapter	
		4. Report/news	
		5. Link/personal blog	
		6. Others	
Cited	B. Type of authorship	1. Single-authored	
		2. Multiple-authored	
Cited	C. Relation to the citing work	1. Reciprocal (self-citation)	
		2. Parallel (cite-coauthor)	
		3. Hierarchical (cite-author with high social capital)	
Cited	D. Location of mentioning	1. Abstract	
		2. Introduction	
		3. Literature Review	
		4. Methodology	
		5. Results/discussion	
		6. Conclusion	
		7. Others (specify)	
Cited	E. Frequency of mentioning	1. Once	
		2. 2 to 4 times	
		3. 5 times or more	
Cited	F. Style of mentioning	1. Not specifically mentioning	
		2. Specifically mentioning but interpreting	
		3. Direct quotation	
Citing	G. Type of citing documents	1. Journal article	
		2. Conference paper	
		3. Book/book chapter	
		4. Report/news	
		5. Link/personal blog	
		6. Other	
Citing	H. Type of authorship	1. Single-authored	
		2. Multiple-authored	
		Semantic (Se)	
Orientation	Categories	Values	
Cited	I. Function of citation	1. Provide background information	
		2. Construct theoretical framework	
		3. Provide previous empirical/experimental evidence	
		4. Describe challenges and limits	
Cited	J. Disposition of citation	1. Positive	
		2. Negative	
		3. Mixed	
		4. Neutral	
Citing	K. Type of research domain	1. Social sciences	
		2. Humanities	
		3. Natural sciences	
		4. Applied sciences and engineering	
Citing	L. Type of research focus	1. Theoretical research	
		2. Empirical research	
		3. Experimental research	
		4. Other (Specify)	

In addition, style of mentioning (Category F) can also indicate the importance of a cited article. Bonzi (1982) argues that a reference that is cited by an article, but is not obviously mentioned in the text, can be considered less relevant than one that is discussed in depth within the text of the citing article. Based on this understanding, we differentiate three styles of mentioning: not specifically mentioning, specifically mentioning but interpreting, and direct quotation. For example, utterances such as “Some studies have proposed . . .,” “For example . . .,” and “e.g . . .” can be coded as “not specifically mentioning” (F1); an utterance such as “Smith (2011) states that . . .” can be coded as “specifically mentioning but interpreting” (F2); an utterance that contains “. . . (Smith, 2011, p. xx)” can be coded as “direct quotation” (F3). It is possible that a reference with a high frequency of Type 3 mentioning is more important, or relevant, than a reference with a low frequency of Type 3 mentioning.

For the semantic module, analysis is usually conducted at the sentence-cluster level to indicate the semantic features of the citation. We have created Categories K (type of research domain) and L (type of research focus) only for citing and not cited articles to identify the potentially different citation patterns across various domains and articles with different research focuses as well as to shed light on the embedded social context.

In terms of citing articles, let us take Paper A as an example. The article is a critical review of historicist and pragmatic theories of information science and was published in the *Journal of the American Society for Information Science and Technology* in 2009. We code Paper A as “K1: Social sciences” because it is published in an information science journal and not a philosophical research journal. It is also coded as “L1: Theoretical research” as it is an understanding and classification of theories of concepts in accordance with epistemological theories (empiricism, rationalism, historicism, and pragmatism). Thus, its main contribution is theory building. Articles that provide conceptual definitions, domain limitations, relationship building, and predictions, offer framework for analysis, facilitate the efficient development of the field, and are needed for the applicability to practical real-world problems fall into this category (Wacker, 1998).

Let us now look at Paper B. It describes the results of a content analysis of the websites of *Fortune* 100 companies and was published in *Corporate Communications: An International Journal* in 2000. We code Paper B as “K1: Social sciences” because it is published in a business journal, which can be included in the general domain of social sciences. Paper B can also be coded as “L2. Empirical research” because it utilized both quantitative and qualitative methods to analyze the acquired empirical data (marketing communications differences across Fortune 100 websites), so as to test the proposed hypotheses and to answer the research questions. According to the *Oxford English Dictionary* (1989), The word *empiric* is derived from the ancient Greek for *experience*. Therefore, empirical

data are based on direct or indirect observations and can be analyzed either quantitatively or qualitatively. Empirical research is any research that generates its findings on empirical data as its test of reality. Such research may also be conducted according to hypothetico-deductive procedures (Fisher, 1959), or de Groot's (1961) empirical cycle (observation–induction–deduction–testing–evaluation).

Paper C describes a systematic, unbiased, and comprehensive approach, termed *interactome capture*, to define the mRNA interactome of proliferating human HeLa cells, published in *Cell*, 2012. It is obvious that Paper C should be coded as “K3: Natural sciences” because it is a research of biological cells and published in a biology journal. In addition, it should be also coded as “L3: Experimental research” not empirical research. Although experimental method is often misunderstood to be equivalent to empirical research, observational studies are not experiments. Experimental research is any research in which data are derived from the systematic manipulation of variables in an experiment (usually a laboratory experiment). Thus, experimental research is more precise and rigid than empirical research in the sense that in an experiment the different “trials” are strictly manipulated so that an inference can be made as to causation of the observed change that results. In general, empirical research adopts a flexible “hypothetico-deductive” (Whewell, 1837/2010) method whereas experimental research is constructed on rigid scientific tests and laboratory work.

In principle, publications in the humanities are usually theoretical works; those in social sciences are often empirical works, and works in the natural sciences and engineering are experimental research. Although exceptions still exist, this principle of connecting “K. Type of research domains” and “L. Type of research focuses” can be used in a computer-assisted coding process.

Function of citation (Category I) is a major measurement to classify cited articles. Instead of presenting too many details, we provide four values based on summarizing previous schemes and extracting the basic research flows: provide background information, construct theoretical framework, provide previous empirical/experimental evidence, and describe challenges and limits. “Provide background information,” which usually appears in the introduction and literature review sections is generated from categories such as “background reading” and “historical” (Duncan, Anderson, & McAleese, 1981), “general informational” and “historical” (Hodges, 1978), “historical background” and “description of other relevant work” (Oppenheim & Renn, 1978). “Construct theoretical framework,” which mainly appears in the methods section, is summarized from categories such as “hypothesis or theory” and “calculation from theory” (Lipetz, 1965), “theory” and “development of ideas” (Duncan et al., 1981), “use of theoretical equation” and “use of methodology” (Oppenheim & Renn, 1978). “Provide previous empirical/experimental evidence” can appear in the literature review, methodology, and results/discussion sections, and is extracted from cat-

egories such as “evidential” (Hodges, 1978), “supplying information or data” (Oppenheim & Renn, 1978), “experimental details” (Duncan et al., 1981). “Describe challenges and limits,” which usually appears in the discussion and conclusion section, is developed from categories such as “questioned” and “refuted” (Lipetz, 1965), “disputing” and “criticism” (Duncan et al., 1981), “oppositional” (Hodges, 1978). These four values are demonstrated based on examples from different scientific domains such as theoretical, empirical, and experimental research. Instances of “I1” (Provide background information) can be identified from the above examples of Paper A, Paper B, and Paper C.

Paper A: “Since philosopher of science Thomas Kuhn (1922–1996) wrote his famous book *The Structure of Scientific Revolutions* (1962), “paradigm” has been a popular term in many fields, although it has also been seriously criticized” (theoretical research).

Paper B: “The number of users of the Internet is estimated at 41 per cent of adults in the USA (Pew Research Center Survey, 1998)” (empirical research).

Paper C: “Taking the natural variation between biological replicates into account, the bioconductor package DESeq (Anders & Huber, 2010) provides a statistical test for assessment of differential abundance of count data” (experimental research).

All the mentioning of previous works offers either historical background (e.g., Kuhn's theory) or information (e.g., facts of Internet use, statistical significance) regarding previous research as an explanation or elaboration of the author's research, no matter whether it is a theory-focused, empirical-focused, or an experimental-focused work. Then examples of “I2” (Construct theoretical framework) are as follows:

Paper A: “(e.g., “formal concept theory” by Priss, 2006) (theoretical research).

Paper B: “Components of the marketing communications mix for Web sites include: advertising, sales promotions, public relations and direct marketing (adapted from Bennett, 1995)” (empirical research).”

Paper C: “Our solution concept is the Perfect Bayesian Equilibrium (PBE) (Fudenberg and Tirole 1991) (experimental research).”

All the above citations represent symbolic concepts, that is, specific terms/concepts (e.g., formal concept theory, components of marketing communication, PBE), which the citing author(s) can borrow, incorporate, and develop to establish the principles and rationales of their own research. In addition, it is obvious that “I3” (Provide previous empirical/experimental evidence) rarely appears in theoretical research. For example, there is no such example in Paper A.

Paper B: “Substantial empirical work has shown that prediction markets produce remarkably accurate forecasts (Berg et al., 2008; Wolfers and Zitzewitz 2004; Goel et al., 2010)” (empirical research).

Paper C: “It has been shown to interact with the 3’ end stem loop of histone mRNA (Yang et al., 2006)” (experimental research).

The citations above refer to the empirical facts that support the citing author’s work. Contextual cueing includes “substantial empirical work” and “it has been shown,” which provide factual evidence or proof. Category “I4” (Describe challenges and limits) is closely related to Category J (Disposition of citation). Although there is an assumption that scientific writing tends to be objective and neutral, there is a distinction between “positive” (acknowledgment) and “negative” (questioning and challenging) citing. For example, citing for establishing an author’s own research foundation, and citing for pinpointing the limits of previous research, can indicate an author’s sentiments:

Paper A: “However, the criticism of Kuhn’s theory of paradigms suggests, among other things, that different ‘paradigms’ do not totally replace each other but exist together and compete with each other in all domains all the time (see, e.g., Mayr, 1997, pp. 98–99)” (theoretical research).

Paper B: “This is a common problem with other Web technologies in which user participation is necessary, for example, recommender systems (Raghavan, 2004)” (empirical research).

Paper C: “Our model is not unique in suffering from a multiplicity of equilibria; multiple equilibria exist in many signaling games as well (e.g. Spence 1973)” (experimental research).

To operationalize this coding, parsing and text mining can be used to identify the cue words such as “however” (Paper A), “but” (Paper A), “problem” (Paper B), “suffer” (Paper C), “nevertheless,” “limit,” “weak,” “undermine,” and “ignore.” Thus, all the above citations can be coded as “I2. Negative” based on these negative cue words. Such a vocabulary can be used for computer-assisted sentiment analysis.

Generally, the codebook we propose provides a relatively comprehensive and balanced framework to conduct CCA. Each citation in the text can be coded and assigned values using Categories A to I, covering dimensions of both citing and cited works, accounting for both individual and collective norms, as well as focusing on syntactic and semantic modules. The expected output is a comprehensive image of citations for different research purposes.

Conclusion and Future Work

Information science researchers have contributed to discussions of scholarly impact and have constructed a sophisticated and widely accepted method to measure it: citation analysis. For example, Voos and Dagaev (1976) suggested that the number of times a reference is cited in an article provides some indication of its relevance to the citing article’s subject. However, Small (1978) also pointed out there is a great deal of evidence that influential articles are more highly cited than uninformative ones, but there is no evidence that highly cited articles are highly influential. In other

words, high number of citations is a necessary, but not sufficient condition of “being influential.”

In addition, citations do not exist in a vacuum, but in an organized scholarly context that also reflects the rich socio-cultural properties, including motivations for citation, functions of citation, sentiments of citations, and so forth. Our goal is not a simplified, one-dimensional citation metrics, but an in-depth, multidimensional demonstration of the epistemological roles played by the citations in the citing article, the heuristic values of the roles played by citations in the citing article (Peritz, 1983), and the interactive network of social/scholarly capital implicated by citations in the citing article. Based on this understanding, we have proposed a framework for a promising method—citation content analysis (CCA), to conduct syntactic and semantic analysis of citation content. We have also provided potential procedures for CCA, including principles to identify the reference scope, a two-dimensional (citing and cited) and two-modular (syntactic and semantic modules) codebook, and possible approaches to operationalize and apply this codebook via computer assistance.

To expand current studies, we will test, modify, and improve the proposed framework and apply CCA to a large-scale dataset, for example, PubMed Central¹ data acquired from the U.S. National Institutes of Health’s National Library of Medicine (NIH/NLM). This test will have both theoretical significance and applied importance.

Theoretically, it can shed light on a few epistemological questions in the current codebook. For example, syntactic features are usually hard to identify for citing articles, and thus only two categories (G and H) are provided in this article. However, we hope that the analysis of the large-scale dataset with the special effort to understand the socio-contextual background of citations can lead to identification of additional syntactic features. Another open question is how to balance deductive and inductive approaches. In this article, we utilized an inductive approach to categorize citation motivation, whereas a deductive approach should also prove to be quite useful to rationalize and model such categorization. For example, some researchers have already discussed the importance of a deductive approach (e.g., Börner, Boyack, Milojević, & Morris, 2012) and applied it to citation studies (e.g., Chen, 2006; Chen & Hicks, 2004; Chen & Yu, 2000). By solving such questions, we hope that our framework will verify the appropriateness of incorporating quantitative and qualitative measures in citation analysis on a large scale. This can lead to a shift within

¹PubMed Central® (PMC) is a free archive of biomedical and life sciences journal literature at the U.S. National Institutes of Health’s National Library of Medicine (NIH/NLM). In keeping with NLM’s legislative mandate to collect and preserve the biomedical literature, PMC serves as a digital counterpart to NLM’s extensive print journal collection. Launched in February 2000, PMC was developed and is managed by NLM’s National Center for Biotechnology Information (NCBI). From <http://www.ncbi.nlm.nih.gov/pmc/about/intro/>

citation analysis from the current purely numerical approaches to richer descriptive and contextual methods, which can provide more details than a simplified one-to-one relationship. Thus, this validation will enrich current citation analysis and open a new frontier for computational linguistics that will focus on understanding and modeling citation patterns, which are quite different from natural language analyses. Apart from the theoretical contribution, the empirical test will result in improved text mining and full-text extraction algorithms, as well as advanced parsing and machine-learning techniques. A possible output may be algorithms and software for intelligently processing language data.

Another possibility is to combine CCA and topic modeling. In scholarly communication, topic modeling is important both at an individual and a collective level. Individually, it is a useful way to mine users' different opinions and attitudes toward various topics. Collectively, it can help analyze the heterogeneous academic domains and networks, facilitating community detecting. Thus, CCA and topic modeling are interdependent and mutually complementary.

In addition, authors' historical citing records can be organized by topic modeling and coded by CCA to map the change of opinions and sentiments these authors had regarding different topics through time, endeavoring to unveil their citing behavior patterns and to detect interrelations between citation motivation and topics. In this way, topic and content similarities can be used to predict authors' citing motivations and opinions on some specific topics, so as to visualize future citing patterns (e.g., stable, increase or decrease). There are at least three hypotheses that can be tested: (a) the authors' opinion is tightly correlated with their topic preference, (b) the authors' opinion is generally shaped in the context of a topic network and thus largely affected by direct influence from peers, and (c) the influence of authors' opinions can also be propagated through their indirect influence through topic network.

Using CCA together with topic modeling, means simultaneously incorporating topic factor and social-sentimental elements in a unified probabilistic framework, which would enable the analysis of the social opinion influence on scholarly networks, and the construction of an interactive influential network of the citing and the cited authors, works, and even patents (Tang et al., 2012). This can be used to detect cross-domain collaborations, which exhibit very different patterns in terms of both content and topics (e.g., sparse connection, complementary expertise, and topic skewness) compared to traditional collaborations in the same domain (Tang, Wu, Sun, & Su, 2012).

Concerning potential research trends in the future, we also suggest an emphasis on the new altmetrics (i.e., "the creation and study of new metrics based on the Social web for analyzing, and informing scholarship"; Priem, Taraborelli, Groth, & Neylon, 2010) that would promote an awareness of the booming social media, and a disposition to interdisciplinary collaboration. With the boom of Web 2.0,

people, including scholars, became inclined to discuss, express, and exchange ideas online. Priem et al. (2010) have pointed out that scholars are increasingly moving their everyday work to the web. For example, online reference managers Zotero and Mendeley each claim to store more than 40 million articles (making them substantially larger than PubMed). The rise of social media, such as Facebook, Twitter, and Microblog (as many as a third of scholars are on Twitter, and a growing number have scholarly blogs), makes forms of scholarly and research impact more diverse than traditional citation metrics. This has led to new challenges for both citation analysis and CA in the field of LIS. Some of the traditional citation analysis methods are hard to apply to these new resources and at the very beginning classical CA was only developed for printed text. With a shared focus on the rich semantic data, CCA and altmetrics can provide potential solutions to the challenges generated by social media, shifting the focus to "how" and "why" from "how many." Altmetrics could track the impact both inside and outside academia, the impact of influential but unofficially cited work (e.g., Twitter mentioning, hashtags, Facebook @), and the impact from sources that are not peer-reviewed. Thus research is needed to correlate CCA, a new generation of citation analysis, and altmetrics, a new generation of citation metrics. This endeavor will combine new tools and existing measures, maintaining traditions while also adapting to new phenomena in the digital age.

In summary, we consider CCA a powerful yet workable way to improve current citation analysis and a necessary supplement for traditional citation metrics. We are interested in understanding the impact of this new approach on analyzing diversified forms of scholarly contribution in today's digital age, and its use in interdisciplinary fields.

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