CITATIONS IN ECONOMICS:
Measurement, Uses and Impacts

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ABSTRACT

I describe and compare sources of data on citations in economics and the statistics derived from them. Constructing data sets of the post-publication citation histories of articles published in the “Top 5” journals in the 1970s and the 2000s, I examine distributions and life cycles of citations, compare citation histories of articles in different sub-specialties in economics and present evidence on the history and heterogeneity of those journals’ impacts and the marginal citation productivity of additional coauthors. I use a new data set of the lifetime citation histories of over 1000 economists from 30 universities to rank economics departments by various measures and to demonstrate the importance of intra- and inter-departmental heterogeneity in productivity. Throughout, the discussion summarizes earlier work, including the impacts of citations on salaries and non-monetary rewards, and how citations reflect judgments about research quality in economics and the importance of economic ideas.
To be occasionally quoted is the only fame I care for.  

[1] [Alexander Smith, 1863, p. 144]

1. Introduction

Why should one care about citations to scholarly articles in economics? Why might looking at citations stem from anything more than narcissistic concerns or reflect behavior among an isolated group of very competitive individuals? Two answers suggest themselves: 1) On a narrow level, within any group of what are essentially independent contractors quality has to be judged. That is true in our business for decisions about tenure, salary and hiring researchers for positions above the entry level. Citations may be better indicators of the quality of a person’s scholarly work than numbers of publications or the kinds of outlets where the research appeared. 2) More broadly, a high enough rate of citation might indicate an impressive career in economics—might say something about a person’s scholarly achievements. One’s citations may be viewed as a reflection of what one has contributed to society in that part of the job that is devoted to research.

We might evaluate the quality of scholars’ contributions by considering where they publish their work, judging its quality by the quality of its outlet. As I show, however, the tremendous heterogeneity of individual contributions published in the same outlet makes attributing journals’ average quality to the individual articles they published extremely error-prone. Alternatively, one might rely on one’s own reading of scholarly contributions, or the readings by a group of scholars refereeing a person for appointment or promotion. This approach may be more sensible, so long as one is willing to discount the tremendous hubris involved in claiming that one’s reading of another scholar’s work provides a definitive evaluation of its quality or if one can rely on groups of contentious economists to offer similar evaluations. Citations, while hardly perfect measures of importance (see Kapeller, 2010, and Wang et al., 2016 for details on problems with citations) are an arguably objective way of judging the research contributions of academics, one that

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1 Although we used this as the epigraph in the first paper on citations in which I was involved (Hamermesh et al., 1982), it is so apropos, since it speaks to the impact of our research on others, that it belongs up front again.

2 Tombazos and Dobra (2014) offer citation-based evidence of self-serving biases in judging economics journals, and presumably too the individual contributions they contain.
is based on what is implicitly a market test, namely whether the scholarly contribution affected the subsequent research of other scholars.

The first purposes of this essay are to illustrate the kinds of data on citations to economic research that are available and to discuss their validity. Having done that, I examine what these data tell us about the nature of citations to scholarly work in our field, looking at differences across sub-fields and journals, both over time and across individual scholars of different ages/cohorts. While I compare individuals and institutions, to reduce the prurient interest in the topic no individuals are mentioned by name, although institutions are.

How to measure citations—what sources are available for this, and how one uses them—is covered in Section 2. Given their availability, what kinds of statistics can be constructed based on these sources? Does it matter what source(s) or statistics one uses to evaluate the quality of published research or of individual authors’ contributions?

Given these measurements, in Section 3 I discuss how citations vary across individuals, sub-fields and journals. Who is a “superstar” scholar in terms of citations? How do citations vary among individuals who differ by experience? More junior people typically have fewer works to be cited (and less time since they were published for them to have been cited), so that I obtain a metric by which to compare the progress of, especially, younger scholars. Focusing still on individuals, I consider differences in citations by gender and by the institutional affiliation of a scholar. The article proceeds to differences in the measured quality of aggregates of economists—the departments of economics in the universities in which they work. In addition to providing a series of possibly novel rankings of institutions, it also sheds light on the extent of heterogeneity within and across institutions.

In Section 3 I also consider differences by article, examining whether resurrection occurs for previously ignored publications, or whether some articles are mere “flashes in the pan.” Focusing still on articles, it considers the marginal productivity, in terms of scholarly impact, of having more authors on a paper, thus estimating a short-run production function for quality. This information is useful for assessing individual scholars’ contributions, an increasingly important issue in light of the now ubiquitous practice
of multiple authorship in economics as compared to practices before 1990 (Hamermesh, 2013). It also examines differences in citations by sub-field of economics, concentrating on methodologies rather than on the decreasingly meaningful distinctions by area of concentration. It considers how differences in citations to articles in different journals have varied over time and examines whether the widespread focus on the “Top 5” journals is sensible. Finally, it points out how different citation practices are in economics compared to other social sciences.

Most of the published research on citations in economics has considered them as measures of quality that might be useful in explaining various outcomes. Chief among these have been salaries (compensation), but several studies have examined other outcomes that are arguably affected by differences in the quality of scholarship. Thus in Section 4 I summarize this body of research, research on citations’ uses in the history of economic thought, and how they have been used to evaluate editors’ and committees’ decisions about publication and awards. The article concludes in Section 5 with a longer apologia for this kind of research and for the enhanced use of citations in judging articles, individuals and institutions; and it offers suggestions for future directions of research relating various outcomes to citations.

Readers should keep in mind that citations in academic journals and books, although important measures of the productivity of an economist’s research, are only partial indicators of its quality. One’s research can, for better or worse, influence broad public attitudes and/or public policy, and it can affect how businesses and individuals organize their activities and make decisions. Moreover, these effects may be direct or indirect, as our research filters through others’ research and through the media. While these additional and often inchoate impacts are likely to be highly correlated with one’s academic citations, I doubt that the correlation is perfect.
2. Measuring Citations and Impacts

2.1 Sources of Information on Citations

Citation analysis in economics has come a long way since Stigler and Friedland (1975) used hand counts through published articles to obtain measures of citations. Today the two most commonly used online methods of acquiring citation counts to a scholarly work or to a person’s works are the Web of Science (WoS), created by the Institute for Scientific Information, and Google Scholar (GS). The WoS [http://apps.webofknowledge.com/] has three sub-categories, for articles in Science, Social Science (SSCI) and Arts and Humanities, as well as two sub-categories for conference proceedings, in Science and in Social Science and Humanities (CPI-SSH). Except for a very few economists whose work has mainly been adopted in more technical fields, obtaining information from the SSCI and the CPI-SSH provides a complete or nearly complete measure of a scholar’s impact. The majority of the studies summarized here use the WoS because they were written before GS became available.

Citation counts in the WoS are based on references included in published articles. Each citation is included under each author’s name, so that a multi-authored article or book will be listed as a citation for each author separately. This was not true when the WoS was only available in print editions (before 2000), but citations to all past articles in the online version now reference all authors. The WoS can also be used to obtain citations to a particular paper by selecting author, year and volume number on the Cited Reference Search screen.

The usual way of using the WoS works from the URL shown in Figure 1. It is based on the cited articles themselves, so that only published articles are included in the citation count. Choosing a search for...

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3Their graduate students performed these tabulations. Given the possibilities for error discussed below when using the online sources now available, I doubt the desirability of using such labor in analyzing citations today.

4Other, apparently less frequently employed (in economics) databases are Microsoft Academic Search and Elsevier Scopus.

5The most extreme case I have encountered is an economist with a meager 407 citations in the WoS whose total count rises to 6658 when the two science indexes are included.
Author and entering the author’s last name and first initial generates a list of all the author’s (journal) publications in descending order of the number of times cited, with citations per year and in total for each, along with additional information shown in Figure 1.⁶

The WoS cites people using the initials of their given names. This method is not a problem for a name like Hamermesh, D (except for misspellings of the surname), but it is a difficulty for someone with a common surname. This problem makes it hard to collect accurate citation counts for some individuals, leading to measurement error of unknown magnitude. The difficulty can be reduced in a search by excluding articles that are obviously classified far outside sub-fields in which an economist might have written, but it is inherent in the WoS.

Searching for citations using GS has the virtue that it is based on references to an article, working paper or book on the internet. It thus has the advantage over the WoS of allowing citations to junior scholars to be available earlier in their careers, which is especially important given the long publication lags in economics that lead to lags in WoS citations. For those economists who have established a GS profile, it provides information like that shown in Figure 2, plus a list of all the referenced publications. If the title, as opposed to the outlet, of an article or book does not change between incarnations as working papers, articles, etc., all references to the piece are accumulated into one aggregate.

The main problem with using GS is that many researchers have not established user profiles, so that for them it is extremely time-consuming to obtain a complete citation count (presumably by searching separately for each entry on their CVs). That difficulty can be partly mitigated by searching for the person in GS, which lists publications arrayed in more or less descending order of the number of citations obtained (and even lists the paper’s citation count in the WoS). Instead of counting citations to all the listed publications, summing those to the person’s five most-cited papers may be a reasonable substitute and is examined in Part B below. Still another serious but much rarer difficulty is that GS will occasionally mis-

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⁶A second method that is used less is based on citing articles and thus includes citations to unpublished working papers, books and other non-journal publications. In this method an article may be listed several times, for example, in its incarnations as working papers disseminated by different organizations and in its published form.
attribute a paper by a scholar with an identical or even similar name as the person of interest, a problem that, as in the WoS, arises most frequently for scholars with frequently occurring surnames.\(^7\)

The problem of identifying names correctly in GS and the similar difficulty in the WoS makes counting citations by web-scraping or having a student who is unfamiliar with the works of particular scholars problematic. Obtaining accurate citation counts is difficult even if one has a feel for who might have written what. Taken together, all of these problems suggest that the most accurate counts by either method will be done by hand by someone familiar with the economics literature (perhaps the researcher her/himself).

Both the WoS and GS provide statistics that seek to aggregate an individual’s citations into one number beyond the total count. The most commonly used of these is the h-statistic, proposed by Hirsch (2005).\(^8\) This statistic is calculated by arraying citations to all of a person’s works from the most- to the least-cited, then defining the statistic as the h-most-cited paper that has h citations. Figure 3 illustrates this calculation. Clearly, this statistic does not measure the mean, variance or any other moment. Instead, it combines both breadth and depth of impact.

As Figure 2 showed, profiles in GS also provide a statistic, i10, indicating the number of a scholar’s papers that have been cited ten or more times. This measure provides a fairly low hurdle, so it can be viewed chiefly as indicating breadth. Other statistics have been proposed for aggregating and ranking citation records. The g-index, offered by Egghe (2006), computes the statistic g such that a scholar’s g most frequently cited publications have \(g^2\) total citations. Clearly, this index will increase when a much-cited publication \(i\) ranked \(g_i < g\) receives an additional citation, while there will be no change in the h-statistic when such a publication, ranked \(h_i < h\), receives an additional citation. Numerous variations on both of

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\(^7\)This difficulty arises especially often in citation counts for Chinese or Korean scholars, given the relative paucity of distinct surnames in those cultures.

\(^8\)Hirsch proposed this statistic as a way of making achievements in various “hard-scientific” fields commensurate, noting that it appeared that Nobel Prizes in these fields were typically awarded to researchers who had attained an h-statistic of 35 in the WoS. An examination of WoS h-statistics among recent Nobel recipients in economics at the time of their award shows huge variations, from below 30 to over 60.
these statistics have been offered, including by Ellison (2013) for a sample of economists. No doubt more can and will be proposed, as recent theoretical work, Perry and Reny (2016) demonstrates; but the h-statistic (in both the WoS and GS) and i10 (in GS for those scholars with profiles) are the only ones readily available in all disciplines, and the former is the only one that has been widely used.9

Yet another source, although one limited to economics, is RePEc (Research Papers in Economics, www.repec.org) which has the virtue similar to GS of including citations to unpublished as well as published articles. As with GS profiles, however, not all authors are registered with RePEc, and unlike GS it does not seem possible to obtain information on articles of unregistered authors.

2.2 Comparing Methods and Constructing Data Sets

Few people would wish to spend time using both the WoS and GS to compare scholars or to construct data sets to examine the impacts of citations on outcomes. Given the differences between them, how much does it matter which source one uses? How much does it matter which statistic describing citation counts one focuses on given one’s choice of source?

To study these and other questions I constructed three new data sets based on the WoS and GS: 1) Citations to articles published in the so-called “Top 5” journals—the American Economic Review, Econometrica, the Journal of Political Economy, the Quarterly Journal of Economics, and the Review of Economic Studies, referred to hereafter as J5. These are viewed by many economists as the most prestigious in the profession, ones that receive the most attention in hiring and promotion decisions.10 This data base covers 1974-75, years arbitrarily chosen to allow a long post-publication period over which citations might accrue. It includes all 444 articles that are at least five pages long in print, and it excludes papers (e.g., presidential and Nobel addresses) that were obviously not refereed or that were clearly comments. 2)
Citations to articles in the 2007 and 2008 volumes of J5, with the same exclusions but limited, given the increased logorrhea in articles in economics journals, to the 497 publications that are at least ten pages long in print. The sample period allows some time for the articles to have had an impact on subsequent scholarship while still being nearly contemporary. I obtained citation counts from both the WoS and GS in this data set. 3) Citations to all 1043 tenure-stream or tenured faculty members with primary appointments in economics departments at the top thirty U.S. economics faculties (ranked based on a history of recent publications in Kalaitzidakis et al., 2003).11 For all of these scholars I obtained information on citations to their five most-cited papers from GS (C5 hereafter). For the 543 people for whom they were available I also obtained entire GS profiles; and for some of them I obtained total WoS citations. Later I examine whether C5 appropriately represents a scholar’s total contribution by comparing it to total citations for the roughly half-sample for which both measures are available. The first two data sets created here contain all citations through December 2014; the third is based on citations cumulated through early May 2015.

Clearly the journals that provide the articles used in constructing the first two samples are an elite small fraction of all the scholarly outlets available to economists. In Section 3.5 I provide some evidence on how moving further down the journal prestige hierarchy might affect the inferences made here, but the main results apply only to research in the most visible outlets in the field. Similarly, results based on the sample of economists from a set of 30 leading institutions will not describe the results one might obtain at other, typically lower-ranked institutions. In Section 3.4, however, I do make a brief excursion to examining comparisons within a broader set of institutions.

Examine first the differences that result from using the different methods by comparing WoS to GS citations to J5 articles published in 2007-08. Taking all but the eleven most heavily (GS) cited papers among the 497 in this sample (to exclude extreme outliers), Figure 4 shows the scatter of the relationship between the two measures for the first six or seven post-publication years. Clearly, the scatter lies quite close to the

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11Faculty who were in what was clearly a “teaching track,” as indicated by titles such as “professor of the practice” or “senior lecturer,” were excluded from this sample. I obtained faculty members’ names and appointments from each department’s website.
regression line (and would look even closer, and the fit would be still tighter, if I included the eleven outliers). A good rule of thumb in extrapolating from GS to WoS citations to an article is that the latter will be about one-fifth of the former.

Since much of the argument for using GS rather than the WoS is the former’s inclusion of citations to unpublished papers, I concentrate on differences between the two sources for more junior faculty. Figure 5 presents a regression of WoS on GS citations based on 102 of the 109 faculty members in the third data set who had GS profiles and who received their Ph.D. degrees after 2005 (again excluding extreme positive outliers). As with the comparison of citations counts by articles, here too it appears not to matter very much which method one uses (and again, were the 7 heavily-cited individuals included, the fit would be even tighter).\(^\text{12}\)

RePEc citation counts are available for 73 percent of the 1043 economists in the third sample (and are reported in Ozarbayev, 2016). The rank correlation of C5 and the RePEc citation count for these authors is 0.94. Even among those scholars whose Ph.D. was awarded after 2005 the rank correlation of the two measures is 0.90 (and that of the RePEc count with the WoS citation count is 0.84). Rankings using RePEc counts are remarkably similar to those using WoS and GS.\(^\text{13}\)

Consider the Spearman rank correlations of various statistics describing citation counts, shown in Table 1 for the 52 percent of scholars in the third data set who had GS profiles. Examining pairwise correlations among \(h\), \(i_{10}\), \(C_5\) and total citations, the rankings are remarkably similar across metrics. Just as it does not matter much what source one uses, it also does not matter much what statistic one uses to measure citations given the source. Since there is no “right” source or “right” statistic, one should aim for breadth and ease of acquisition. With GS profiles only available for barely half the researchers at top

\(^{12}\)Restricting the sample still further to those who received their doctoral degree after 2006, or even after 2007 or 2008, does not qualitatively alter the slope of the regression or the adjusted \(R^2\).

\(^{13}\)The likelihood of having a GS profile (RePEc registration) is significantly greater among those who, at the same post-Ph.D. experience, had a higher \(C_5\) and among men. It rises significantly as experience increases up through 12 (20) years and falls significantly thereafter. RePEc citations are about 1/8 of GS citations among those who have both a GS profile and a RePEc registration.
economics departments, these considerations suggest using C5 when comparing individuals or using citation data to predict outcomes. C5 is very highly correlated with total citations, suggesting that it is an excellent proxy for total citations for individuals without GS profiles. It is also very highly correlated with citations to a person’s three most heavily-cited works (rank correlation = 0.998, with rank correlations of 0.996 within sub-samples of scholars with 30+ years or fewer than 10 years of experience). In the end, for papers published in J5, and for individuals located in top economics departments, one’s choice of source and statistic is unlikely to affect greatly any conclusions or rankings. Finally, the rank correlation of C5 with WoS citations for scholars with a post-2005 Ph.D. is 0.96, paralleling results in Henrekson and Waldenström (2011) in a small sample of Swedish full professors.

One might worry that citations are phony, in the sense that citation counts are swollen by negative citations (criticizing a work) or by references by members of one’s scholarly network (students, colleagues, etc.). Undoubtedly we do cite our colleagues, coauthors and mentors, especially when we are new to the field (Clements and Wang, 2003). Cole and Cole (1973), however, demonstrate for the physical sciences that the correlation between all of one’s citations and the total excluding what are viewed as negative citations is very high. This result is not surprising: Scholars usually do not waste time attacking research that has not already garnered attention from other scholars. Similarly, and based on the RePEc data constructed by Ozarbayev (2016), excluding citations by one’s coauthors and members of a slightly more broadly defined network hardly alters the rankings of citations to the scholars in the third sample used here.

This essay deals with citations in economics; but, as noted above, the justification for using citations is that they are a market-based measure of the quality of research. They are by far the most widely examined measure; but downloads of articles also measure other scholars’ interest in one’s work. To examine the relationship between downloads and citations, I obtained the total number of downloads for the sub-sample of scholars in Top 30 departments who were registered on RePEc. Unlike in GS, books will not be

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14 With the sample limited to 30 very high-quality departments I cannot be sure that the same results would hold if I had cast the quality net more widely. That the rank correlation of 0.982 between total and C5 citations shown for this sample in Table 1 rises to 0.985 in the lower-quality half-sample in this group does, however, suggest that the result holds fairly generally.
downloadable and cannot be included in these statistics. Nonetheless, for those scholars whose Ph.D. was obtained after 2005 the rank correlation between C5 and the number of downloads was 0.78; and for those who received their Ph.D. in 1980 or earlier the rank correlation was 0.83. Clearly, downloads provide very similar information on research quality to citations.

3. Differences in Citations across Individuals, Articles, Sub-Fields, Institutions and Journals

In this Section I consider a wide range of issues focusing solely on the bibliometrics of citations—on their variation across a wide range of indicators. A central fact runs through all these comparisons: The distributions of citation measures are highly right-skewed. For that reason, throughout this Section I present data describing the shapes of the distributions of citations, not merely measures of central tendency, particularly means, which have been the main focus in the literature.

3.1 Differences among Individuals

Within the sample of 1043 economists from elite institutions there are tremendous differences in the extent to which their works have been cited and have thus presumably influenced the work of other scholars. As Table 2 shows, the mean citations to the five most-cited papers is about 3000; but the median is less than 1100. The mean h-statistic (in the positively-selected subsample that has established GS profiles) is 25, while the median is 17, implying less skewness in this summary measure than in the C5 measure or, as the table shows, in the counts of total citations in this sub-sample.

While the term “superstar” is widely and loosely applied, one might ask what constitutes a “super-economist?” Consider the ten scholars whose citations place them in the top 1 percent of the distributions shown in Table 2, Column (1). Their average C5=38,818, arguably making them true super-economists. Clearly, these are the most influential people in this profession in the United States, at least as measured by the impact of their research on that of other scholars. Moreover, given the pre-eminence of U.S. universities
in economics, and the limitation of the sample to what are arguably the top 30 U.S. economics departments, these scholars stand on an even higher pinnacle in the profession worldwide.\textsuperscript{15}

The citation measures in this sample reflect cumulative citations over a scholar’s publishing lifetime. Other things equal, this generates a positive citations-age profile, so that the raw data described in Table 2 tend to place more senior scholars higher in the distributions of the various measures. This tendency may be offset in the GS measures by their inclusion of references to discussion/working papers, series of which and their availability have proliferated with the creation and expansion of the internet. The availability of online indexes may also have generated a positive trend in citations per article (McCabe and Snyder, 2015). It is also more than offset in the WoS by the growth in the numbers of sources included in that data set.\textsuperscript{16}

I examine the life cycle of citations (cross-section data, and thus not indicative of the life cycle of citations of a particular cohort) by including for each individual in the sample his/her years of post-Ph.D. experience (2015 – Ph.D. year), \textit{a}. This information might be useful for comparing younger scholars who are being considered for re-appointment or for tenure, and perhaps even for considering the achievements of more senior scholars who might be appointed to new positions. Table 3 presents these citation-experience profiles, aggregating more experienced individuals into broader categories. Perhaps most noteworthy in these data is that the skewness demonstrated above in the aggregate persists as experience increases and, indeed, remains roughly constant (comparing the ratios of means to medians).

\textsuperscript{15}For examples, the most-cited academic economist in the United Kingdom would be in the 99\textsuperscript{th} percentile of scholars, as ranked by C5, if in the U.S. sample; the most-cited in the French-speaking world would barely be at the 99\textsuperscript{th} percentile; the most-cited economist in the German-speaking part of Europe would “only” be in the 98\textsuperscript{th} percentile; the most-cited economists based in Canada or Israel would only be at the 97\textsuperscript{th} percentile in this ranking; and the most-cited economist in Australia would be at the 96\textsuperscript{th} percentile.

\textsuperscript{16}Without longitudinal data one cannot distinguish among trends in citations, individuals’ life-cycles of citations, and differences in citations across cohorts (as in Borjas, 1985, describing immigrants’ earnings). Using an unbalanced panel of WoS data describing citations from 1991-2013 to 111 high-achieving labor economists whose Ph.D. degrees were received between 1955 and 2002 (clearly not a random sample), one finds that at the same age (or cohort—we cannot distinguish between them) there was a near doubling of the number of WoS citations over the 22 years in the sample. Alternatively, examining the history of citations to articles published in the \textit{American Economic Review}, Margo (2011) finds roughly a quintupling of early post-publication citations to them in six elite economics journals (J5 plus \textit{Review of Economics and Statistics}) between 1960 and 2000.
These are not longitudinal data, and following individuals’ citations through their careers is extremely cumbersome. Some inkling about their time paths is easily obtained for those scholars with GS profiles. Estimating the regression of GS citations from 2010 through May 2015 on lifetime GS citations among scholars with at least 31 years of experience yields a coefficient of 0.39, with an adjusted R² of 0.98.¹⁷ A least-absolute deviations (LAD) regression yields the identical slope, although the pseudo-R² is “only” 0.85. Despite the skewness in citations in all the samples used, parameter estimates generally differ very little between least-squares (OLS) and LAD regressions, a result that persists in all the relationships estimated in this study.

One could use this information *mutatis mutandis* to compare scholars within and across cohorts. Let \( C_i \) and \( C_j \) be the citation counts of scholars \( i \) and \( j \), and let \( a_i > a_j \) be their (Ph.D.) ages. Then if \( F(a) \) is the cross-section citations-age profile, under the assumption that \( F \) is independent of the (constant) secular annual rate of growth of citations per article, \( \theta \), the appropriate comparison of \( i \)’s and \( j \)’s scholarly impact is:

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(1) \quad C_i - F(a_i) - \theta [a_i-a_j] \text{ to } C_j - F(a_j).
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One might be interested in extending this calculation, if one could determine average differences in citation practices across sub-fields (see below), to compare scholars in different sub-fields in economics.

The data set also allows one to examine whether there are gender differences in the extent to which scholars are cited, a claim that was made and disputed in research from the 1980s (Ferber, 1988; Laband, 1987). Finding a differential rate of citation to male and female authors would tell us nothing about the presence of discrimination (just as demonstrating the existence of a wage differential by gender cannot inform us whether it arises from discrimination or differences in productivity), unless one assumes that

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¹⁷Lifetime total GS citations in this sample of 108 very senior scholars range from 593 to 214,046. Even with the possible selection generated by less successful researchers leaving university jobs at earlier ages, the data still cover a very broad range of scholarly recognition.

¹⁸I am indebted to Jeffrey Frankel for suggesting this exercise. If the two underlying assumptions are violated, a more complex formula can be constructed fairly easily with some additional work.
published papers are inherently of equal quality across gender. I thus present an outcome, saying nothing about causation.

To examine these issues I expand LAD regressions that relate C5 to the vector of indicators of experience by adding an indicator of gender (female). I estimate these equations separately for half sub-samples arrayed by post-Ph.D. experience (with the median experience being 17 years). In the older half-sample only 9 percent of the faculty are female, while in the younger half 19 percent are. The estimates, shown in Columns (1) and (3) of Table 4, suggest that older female faculty have indeed been cited less than their male contemporaries, but the difference is small and statistically insignificant; and the difference is essentially zero among younger faculty. When we add the rank of the economist’s department (1 being the department whose median-cited faculty member has the most citations), to examine whether female economists’ citation totals differ from those of men in equal average-quality departments, the results, shown in Columns (2) and (4) suggest little change in the conclusions. There may have been gender discrimination in citation practices in economics in the past, but this evidence suggests that it was small, and it is not apparent in the treatment of younger economists.

Several studies (e.g., Eiran and Yariv, 2006; Huang, 2015) have argued that the position of one’s name in the alphabet affects success in the economics profession, such as the attainment of a tenured position, perhaps due to the practice of listing coauthors’ names alphabetically (Engers et al., 1999). This may have been true before the growth of the internet, when hard-copy volumes of the WoS listed only first authors, but it seems less likely to be true now that all authors are included in citation counts. To examine whether there are alphabetical effects on citations, the names of the authors in the sample were alphabetized and a variable ranging from 1 to 1043 was added to the LAD estimating equation. Moving from the first to the median (in the alphabet) author reduces C5 citations in the more senior half-sample by 1.8 percent ($t = -0.55$), and among their younger colleagues it reduces citations by 0.7 percent ($t = -0.22$). Those whose surnames are late in the alphabet are cited less frequently than their “A” colleagues, but the differences are slight, statistically unimportant and, as expected, smaller in percentage terms among more junior scholars.
Another difference that may be viewed as a kind of discrimination might be the existence of bandwagon effects in citations: One paper among several similar ones is cited shortly after it appears, and subsequent papers disproportionately cite it rather than equally important papers that appeared at the same time—or even earlier. Partly this sort of herding may reflect what Merton (1968) identified as “the Matthew Effect”—the attribution of credit to the better-known scholar in a pair or even in a group of scholars who have contemporaneously produced very similar work.\(^9\) No doubt this kind of behavior occurs and is unfortunate (for lesser-known people), but its extent is completely unclear.

One might worry that some citations are self-citations and do not reflect an article’s true scholarly productivity. One study of articles in the AER and the Economic Journal found WoS self-citation rates, which I denote as the “Ego Index,” approaching ten percent (Hudson, 2007). The extent of self-citation is impossible to obtain from GS, but WoS citations are summarized in total and without self-citations. Rankings among individuals or articles are hardly altered when self-citations are excluded. Selecting the junior economists (Ph.D. year after 2005) in the sample of 1043 faculty, the rank correlation of total citations and citations by others is 0.99, with an Ego Index having a mean of 0.07 (median of 0.05, range 0 to 1). Randomly selecting 100 of the senior (31+ years of experience), the rank correlation of total citations and citations by others is also 0.99, with an Ego Index having mean and median of 0.03 (range of 0 to 0.16). Moreover, additional self-citations do not generate more citations by other scholars (Medoff, 2006).\(^{20}\)

3.2 Differences among Articles

Just as the distribution of citations across a group of scholars is highly right-skewed, so too is the distribution of citations across articles, including those in J5. In the sample of 444 articles published in these journals in 1974 and 1975, total SSCI citations from 1974 through 2014 ranged from 0 to 2466, with

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\(^9\)I was the “victim” of the Matthew Effect when a visiting lecturer referred in his paper and presentation to “Gary Becker’s important work on suicide.” I was not in the audience, but one of my colleagues pointed out that it was my work, and that Becker had never written on the topic.

\(^{20}\)Frances Hamermesh noted that this article may be aimed at adding to my citation count (even though its Ego Index is “only” 0.07). One study in the 2007-8 sample of articles cited 72 works, of which 20 were by the sole author of the paper (Ego Index of 0.28). A four-authored article in that sample had a collective Ego Index (an oxymoron) of 0.40.
a mean (median) of 75 (22); and even the paper at the 95th percentile received “only” 360 citations over its forty post-publication years. Only five percent of these elite articles were uncited (compared to 26 percent of articles in a much broader set of journals examined by Laband and Tollison, 2003). The extent of skewness is illustrated by the distribution graphed in Figure 6a, in which, to make citations across journals commensurate, I have standardized articles by page size (using AER2008 pages) and divided the citations of each article by its AER2008-equivalent pages.21 As the figure shows, most papers, even in these prestigious outlets, are very rarely if ever cited, with relatively few articles accounting for the attention that scholars concentrate on these journals.

Of course, we cannot be sure that such skewness will prevail over the future lifetimes of recently published articles; but taking the 2007-08 sample, the distribution of citations (through 2014) per AER2008-equivalent page, shown in Figure 6b, looks very much like that in Figure 6a. The statistics describing the distribution are also similar, with total SSCI citations ranging from 0 to 478 and a mean (median) of 50 (36); and even the paper at the 95th percentile received “only” 142 citations in its first seven post-publication years. Only one of these articles was uncited, perhaps reflecting the growth of the literature, perhaps indicating an increase in the focus on J5 by authors publishing elsewhere.

The distribution of citations to more recent articles exhibits less skewness than that to the older articles over their longer post-publication lifetimes. Has there been a secular broadening of interest in papers published in these journals, or do articles that are little-cited early post-publication die off more quickly than those that have become “superstar articles” in the eyes of future scholars? Taking the first ten years of post-publication WoS citations to articles in the 1974-75 sample, the mean (median) citations are 22 (11), with a range of 0 to 333, and “only” 82 citations at the 95th percentile. Like the distribution of citations to 2007-08 articles, the distribution of initial (first-decade) citations to 1974-75 articles exhibits less skewness than the distribution of citations over a much longer period. “Star” articles become superstar articles, while “asteroid” articles become dust.

21In this sample each additional AER2008-equivalent page generated an extra 7 citations over the 40 or 41 years after the article’s publication.
This evidence suggests that there generally are few “flashes in the pan”—papers that attract much attention initially and are then ignored—and few “resurrections”—papers that are ignored and come to life much later—among scholarly articles in economics. Perhaps, however, there are at least a few exceptions—perhaps there is hope for articles that were once ignored, and perhaps some that receive substantial early attention are soon ignored. Table 5 compares citations in the post-publication years 21-40 of the 1974-75 articles to their ranks in citation counts in their first ten post-publication years. None of the twenty percent of initially least-cited papers is among the top twenty percent in citations in the most recent twenty years; and most remain ignored. Obversely, while a few of the papers that are in the top fifth of citations in the first ten post-publication years are eventually ignored, most remain well-cited even in post-publication “old age.” There are very few flashes in the pan or resurrections, even among articles published in the most visible outlets.

The mean citation in this sample (among the 423 articles that received at least one citation) occurred 14 years after the article’s publication. Consistent with the evidence above, however, the mean age of citations to the 40 percent of articles with the fewest (nonzero) citations was 11 years, while that to the 17 percent with the most citations (at least 100) was 20 years. The initially most-cited papers “last longer” than others.22 One’s publications do not last as long, however, if one is no longer active: Aizenman and Kletzer (2011) provide stark evidence of this, showing that citations to articles by very productive economists who died young drop below those to contemporaneous articles written by otherwise identical contemporaries who survived them. And explicit citations to the work even of historical superstars eventually diminish (Anderson et al., 1989); their work is eventually either ignored or becomes part of the background that today’s scholars take for granted.

How long a citation history might be needed to predict the long-run impact of a scholarly economics article? In a negative binomial regression of citations to articles in the 1974-75 sample, citations in post-

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publication years 11 through 40 are significantly related to citations in each of the first and second five post-publication years. The relationship is, however, much stronger for the citations in the second five years. This suggests that using the first few years of scholarly interest in an article is not sufficient to gauge its long-term impact, and that a regrettably longer period is required. Nonetheless, faute de plus even using five years of citations provides some useful information about long-term impacts.

With the entire distribution of the number of authors listed on an article in economics moving steadily rightward, an important issue in evaluating people’s records for appointment or promotion is how to account for multiple authorship. Even ignoring the possibility that one’s coauthors are surely more likely to cite their own works, which will increase one’s citation count the more coauthors are listed on an article, does having more authors on a paper add to its scholarly impact? What should be the divisor of citations when evaluating an article, with extremes being 1 and \( N \), the number of authors?

In the very long run the answer to this question should be: Divide by \( N \), since eventually the full impact of an article has become known and, absent the kinds of specific \textit{ex ante} attributions of credit that appear in many published articles in the hard sciences, credit should be divided equally. This rule is of little practical help, since the evaluation of research proceeds in the very short-run. One can at least get a hint of what the long-run production function might look like, however, by examining citations in various intermediate runs. Consider an article-based production function of quality as:

\begin{equation}
Q_i = Q(N_i, T_i),
\end{equation}

where \( i \) is an article, \( Q \) can be proxied by citations, \( N \) is the number of authors, and \( T \) is the technology of producing research of a given quality at time \( t \).\footnote{Flores-Szwagrzak and Treibich (2016) propose a person-based adjustment, combining the citation counts to a scholar’s sole- and co-authored papers.} This relationship has been examined in several studies using specific samples, but not comparing publications in two widely-separated time periods, and not concentrating on J5 publications. We could estimate regressions describing \( Q_N \) and even infer something about \( Q_{NT} \), thus whether the expansion path of \( Q \) in \( N \) is homothetic in changes in technology. The size of
and the sign of $Q_{NT}$ can, however, be inferred simply from an examination of the means and centile points of the distributions of WoS citations as functions of the number of authors listed on the paper.

Table 6 presents these statistics. There is no significant difference at any point of the distribution of lifetime citations between the one-fourth of papers that were multi-authored in the early period and those to single-authored papers. In the more recent period, however, multi-authored papers did receive more citations in their first seven post-publication years, a difference that is visible across the distributions of citations. But going from one to four or more authors roughly only doubles the number of citations at each point of the distribution—it does not quadruple it. Moreover, since additional authors lengthen papers in these samples, if we use citations per page the differences by number of authors are even smaller.\textsuperscript{24} Taken together, the results suggest that today $Q_N > 0$ and $Q_{NN} < 0$, and indicate that $Q_{NT} > 0$. These findings are consistent with those of Hollis (2001), who measured production by the impact factors of the journals where the articles appear, and Medoff (2003), who used citations to individual articles published in 1990. Having more coauthors on one’s papers may induce additional citations to one’s other works, partly since coauthors become familiar with papers that they may feel obligated to cite (Bosquet and Combes, 2013), but the effect of additional authors on a paper’s citations is clearly not proportional to the number of authors.

Given the impossibility of evaluating the long-term impact of an article and dividing by the number of authors $N$, a good rule of thumb suggested by the results in the bottom part of Table 6 is that job candidates’ citations—their influence on other scholars—might well be divided by $N/2$ for $N>1$. At the very least, one should discount citations to multi-authored articles by some factor.\textsuperscript{25}

\textsuperscript{24}The results tabulated in the upper part of Table 6 do not change qualitatively if we restrict citations to the first seven post-publication years, thus making them comparable to those in the later sample. They also do not change very much if we use GS rather than WoS citations to describe the production function in the later sample.

\textsuperscript{25}There is also some evidence that having additional authors does not help get papers into more prestigious journals. The mean number of authors of articles in the later J5 sample was 2.01. The mean number of authors of articles published in 2007-08 in a pair of general-interest journals that might be viewed as the next rung down the journal prestige hierarchy, the \textit{Economic Journal} and the \textit{Review of Economics and Statistics}, was 1.97, not significantly less.
3.3 Differences among Sub-Fields

For over 50 years the JEL and its predecessor, the Index of Economic Abstracts, have maintained a (changing) coding system for classifying journal articles. At a time when macroeconomic research is increasingly micro-based, and when applied work in such formerly diverse fields as economic development, labor economics, public finance, and perhaps others seems increasingly characterized by similar approaches, it may make sense for purposes of analysis to use a methodology- rather than a subject-based classification of published research. This approach currently requires hands-on examination of each paper rather than an easy reliance on author-provided codes, but it may be more appropriate given the changes that have occurred in the field over time. (See Cherrier, 2015, and Kosnik, 2016, for discussions of various difficulties with the JEL coding system.)

I thus classify the articles in the 2007-08 data set into the following six categories: Theory; theory with simulation (including the substantial part of recent macroeconomic research that uses calibration methods); empirical with borrowed data; empirical with own data; experimental; and econometric theory.26 Because no experimental studies are included in the 1974-75 data set, and very little empirical research was based on data sets that the author(s) had created, for the earlier sample I divide publications into those in theory (including theory articles that also involved simulations), empirical economics and econometric theory.

Statistics in the upper two panels of Table 7 describing the two data sets based on J5 articles show that in their first seven post-publication years empirical and experimental studies generated more citations than did articles in economic theory or econometric theory in these journals. (This conclusion is similar to that of Anauati et al., 2016, who used GS citations and a somewhat similar classification of articles.) This is true for the average paper, true at the medians and even true at the 90th percentile. The only caveat to the

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26Excluding econometric theory, because it did not include Econometrica, the same classification was used by Hamermesh (2013).
conclusion that empirical research generates more subsequent attention than theoretical work is at the upper extreme of citations: The most heavily-cited study in the earlier sample was in economic theory.27

One might argue that this comparison ignores the long-lasting effect of theoretical work (a point similar to that made by Chiappori and Levitt, 2003), and that a fair examination of the attention paid to articles based on different methodologies would cover a longer period of time. To examine this hypothesis for the earlier data set, thus allowing forty years of post-publication citations to accumulate, modifies the conclusion based on the first seven post-publication years of citations only slightly. At the median and 90th percentiles of the distributions, empirical articles dominate articles in the other categories; but because four theory articles are the most cited in this sample, there is no significant difference at the mean between lifetime citations to articles in economic theory and those in empirical economics. Since in the later sample the upper tail of the distribution of citations to empirical studies was denser than that to theoretical studies, there may be some doubt whether the long-run impact of theory articles, as measured by citations, will be as relatively large among recently published articles as it was in the past.28

Yet another possibility is that there are differences across sub-fields in the amount of citing, which, if citations are disproportionately to other research in the same sub-field, could generate the results shown in Table 7. To examine this possibility I counted the number of items cited (excluding data sources) in the sample of articles from 2007-08. Across the six sub-fields listed in the table, the median numbers of cited items per article were: Theory, 32; theory with simulation, 37; borrowed-data empirical, 39; own-data empirical, 35; experimental, 33, and econometric theory, 38. These differences seem fairly small; and the extent to which citations are concentrated within sub-fields is, however, unclear, as many of the theory

OLS and LAD estimates that include indicators for the journal where the article was published yield inferences very much like those reflected in the statistics in this table.

Citations to even the most heavily-cited article in econometric theory in the later sample were surpassed by those to seven empirical articles, suggesting that the relative long-run impact of recent empirical studies and those in econometric theory will not differ from their short-run impact.
papers motivate themselves by references to empirical findings, and many of the empirical studies cite theoretical work.\textsuperscript{29}

Even though the quantities of citations do not appear to differ greatly across sub-fields, ideally in comparing individuals one would like to account for differences in citation practices across the sub-fields in which they work. The difficulty, of course, is that many economists work in several sub-fields, whether classified by major JEL category or by the six-fold classification used here. Perhaps the best conclusion is that this consideration should be an important caveat in any comparisons of individual scholars’ productivity.

A related problem is that publication rates by sub-field, and thus to some extent the possibility of generating citations, may differ across sub-fields. Some attempts have been made to account for this possibility, normalizing citations by sub-field (as van Leeuwen and Calero Medina, 2012, and Bornmann and Wohlrabe, 2016, have done using JEL codes). Ignoring the issue of the JEL codes’ appropriateness as sensible classifications of sub-fields, one might argue that normalization by sub-field may not make much sense. If a sub-field is tiny and of little interest to most economists, having a relatively highly-cited article within the sub-field is arguably less important than a relatively less well-cited article in a sub-field of greater interest to most scholars.

3.4 Differences Among Institutions

Ranking economics departments by quality is a long-standing enterprise. See, e.g., Cartter (1966), Roose and Anderson (1970) and National Research Council (1995), based on surveys of impressions (with Thursby, 2000, showing that ratings only partly reflect publication and citation counts of the rated departments); Gerrity and McKenzie (1978), perhaps the first article to use citations in economics, and Davis and Papanek (1984) more broadly, based on average citations; Graves \textit{et al.} (1982) and many

\textsuperscript{29}Still another possibility is that, despite the general evidence of the similarity of patterns of WoS and GS citations shown in Figures 4 and 5, the relationships between the two measures differ across sub-fields. Holding constant indicators of the journals where the articles in the 2007-08 sample appeared, the only significant differences across sub-fields are that both experimental work and econometric theory receive more relatively more WoS than GS citations compared to those received by articles in the other four sub-fields.
followers, based on pages published in selected sets of journals; Tschirhart (1989) and Dusansky and Vernon (1998), based upon measures of the average quality of the journals in which faculty members published; Amir and Knauff (2008), using a measure of the quality of placements of Ph.D. students; Mixon and Upadhyaya (2012), based on the awards faculty received; and Ellison (2013), based on average h-statistics and other transformations of citation counts. I produce yet another such ranking, using the data on the 1043 scholars in the 30 economics departments. The novelty here is several-fold: 1) I examine differences in rankings of institutions based on comparisons at various quantiles of the distributions of their faculties’ citations (rather than at the means), which is important given the tremendous skewness of citation counts demonstrated in Table 3; 2) I measure the extent to which the quality of faculties, as indicated by citations, overlaps across departments; and 3) I adjust for inter-departmental differences in the distributions of seniority.

Authors of any study that compares departments by citation counts must decide on several issues: 1) Use the count of citations to articles/books ever published by the current staff, or use citation counts based on authors’ affiliations at the times of publication? The former approach advantages institutions that recruit much of their staff to senior positions, the latter helps schools that might be viewed as “training grounds.” 2) Use lifetime or recent citations? 3) Use order statistics or averages of citations? 4) Use individual-based statistics or departmental totals (thus advantaging larger faculties)? There is no correct answer to these questions, as they depend on one’s view of what determines the scholarly reputation of an institution. In what follows I choose the first answer to each question. I do provide enough information, however, that one can construct rankings based on the alternatives to the third and fourth choices.

I first calculate $r_{d50}$, the rank of the department by the citations of the median-cited faculty member, $C_{50}$. I then also calculate $r_{d50} - x^*$, the highest-ranked department in which, if they were transferred there, the top quartile of faculty members (ranked by $C_5$) in Department d would be above the median faculty member’s citation count in that department.
The results of these calculations are presented in Table 8. The first column lists the number of faculty members included in each department, while the second shows the department’s ranking, \( r_{50} \) and the third presents \( C_{50} \). The median citations in the top-ranked department are an extreme outlier from the rest of the sample; beyond that, the differences in \( C_{50} \) among departments are much smaller over the broad range from the 5th through the 14th-ranked department.

The fourth column shows rankings in the study from which this sample of 30 departments was taken (Kalaitzidakis et al. (KMS), 2003). Even though \( r_{50} \) is based on lifetime citations to articles ever published by the median-cited person in an economics department in 2014-15, while KMS used journals’ impact factors and total impact-factor weighted publications of articles published in 1995-99, the correlation between these rankings that differ greatly in methodology and timing is 0.70.

The apparent inter-institutional diversity in quality, however, does not reflect the tremendous intra-institutional diversity that underlies these statistics. The calculations of \( x^* \) in the fifth column show that the scholars in the top quartiles of schools in all but three of the departments ranked 11 through 20 are more heavily-cited than the median scholar in at least the tenth-ranked department; and those in the top quartile of economists in schools ranked 6 through 10 are more heavily-cited than the median scholar in the school ranked second. The top people in lower-ranked institutions (in this elite group) would, if the correlation of departmental and individual rankings were one, be placed far higher than their current employment indicates. The overlap is consistent with evidence from Kim et al. (2009) of a decline in the effect of one’s

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30The rankings in KMS are necessarily higher in larger departments; and, as they include all publishers from an institution, not only those in an economics department, an institution ranked more highly if it contained more people outside an economics department who published in economics journals.

31The rank correlation between the medians measure shown in the Table and the mean citations across departments is 0.87. Listing the ranks of schools by mean citations in the same order as in the Table yields: 1, 3, 4, 6, 2, 5, 8, 7, 19, 9, 11, 21, 16, 12, 10, 18, 15, 17, 20, 26, 22, 28, 14, 25, 13, 24, 27, 29, 23, 30.

32With one exception, the replacement of Yale by NYU, the top seven schools in this ranking are the same seven viewed as distinguished in the Carter (1966) ranking, which was based on a survey fielded in 1964. Carter’s compendium sought to be a nationally representative set of opinions, following earlier studies that were viewed as parochial (Hughes, 1925, and Keniston, 1959).
academic affiliation—the quality of one’s colleagues—on one’s scholarly productivity, although obviously the Table does not provide longitudinal evidence.

The most productive (top quartile) economists in a large number of institutions would raise the median quality of the faculties in departments ranked far above theirs. This overlap suggests that newly-minted Ph.D.s might be a bit less concerned about the average quality of departments in which they are seeking or being offered employment. The same comment applies to university graduates applying to Ph.D. programs: There are large numbers of high-quality potential mentors at many institutions (and there is convincing evidence of the greater relative importance of the scholarly achievements of a Ph.D. student’s mentor than of the average faculty member at the student’s institution, Hilmer and Hilmer, 2009).

One might argue that faculties should be judged by the achievements of their stars, not by those of their median faculty members. While the Table demonstrates the extent of overlaps in quality, alternative rankings of departments based on single statistics measuring faculty quality in the upper tail do not hugely change one’s perception of distinctions among departments. The sixth column in Table 8 ranks departments by the citations of the faculty member at the 90th percentile of this measure within departments. Even using this extreme ranking 7 of the top 10 departments ranked by median quality remain in the top 10, and 16 of the top 20 ranked by median quality remain in the top 20. The correlation of the two rankings is 0.68 (rising to 0.78 if the Los Angeles–area outliers, two of the smallest departments in the sample, are deleted).

As suggested by Table 3, those institutions with more junior faculty will exhibit lower average and median citations. To account for such differences I generate predicted median citations for each individual i, $C_{5i}^*$, by estimating a LAD regression describing each scholar’s citations $C_{5i}$ as a function of the large vector of indicators of post-Ph.D. experience, $a$, shown in Table 3. I then use these predictions to adjust the citations of each faculty member in department d for experience by taking the residual $C_{5i} - C_{5i}^*$, and then rank departments by the median of this residual to obtain $r_{d50}^*$. The results of this re-ranking are presented in the final column in Table 8. The correlation of this measure, $r_{d50}^*$, with $r_{d50}$ is 0.78. While the adjustment matters, except for a few institutions (those with unusually junior or senior faculties), it does not change
the rankings very much (with this re-ranking particularly advantaging Cornell and UT-Austin, and disadvantage Minnesota).

The sample was constructed using faculty members located in each university’s economics department, so that the rankings necessarily ignore the facts that some institutions have large and often highly productive groups of economists located in other units on campus (e.g., business, public policy, education or law schools) and that the importance of this phenomenon differs across institutions. Since it is impossible to identify all units where economists might be located in each of these institutions, I leave it to readers to make their own mental adjustments of these rankings to account for these inter-institutional differences.

This discussion has concentrated on the top-ranked American economics faculties, perhaps appropriately given the concentration of articles in leading journals written by U.S.-based scholars (Hodgson and Rothman, 1999), a dominance that citation analysis shows is not diminishing (Drèze and Estevan, 2007). Given the increasing flows of economists to and mostly from Europe (and to a much lesser extent, to and mostly from Asia and Australia), however, one might wonder where the top-ranked schools in other English-speaking countries would fit into the rankings presented in Table 8. To characterize this issue simply, I examined median C5 citations of faculty members in the institution in the United Kingdom that received the highest ranking in the 2014 Research Excellence Framework (REF), University College London (UCL), computing the same measures for it that underlay the rankings in Table 8. By this measure UCL would have ranked twenty-first in Table 8; and, as with the faculties in the U.S., its faculty members are very diverse, so that \( x^* = 14 \). A similar exercise for what RePEc rated in 2016 as the leading economics faculty in Canada, that of the University of British Columbia, would have it ranked 26th in Table 8, with \( x^* = 16 \).

These and the many other institutional rankings concentrate on an elite group of research-oriented institutions. On average members of these faculties are more heavily cited than economists at lower-ranked research institutions; but just as within this set of 30 schools, there is substantial overlap of them with faculties at lower-ranked schools. Taking a randomly chosen set of 5 economics faculties in departments
ranked between 31 and 55 in the U.S. in KMS, 32 percent of scholars in those departments were more heavily cited (C5) than the median scholar in the elite sample. Taking another randomly chosen set of 5 faculties in departments ranked between 56 and 90 in the U.S. in KMS, 13 percent of scholars were more heavily cited (C5) than the median scholar in the elite sample. These comparisons illustrate that the extent of overlap diminishes but still remains substantial as one moves further beyond the most elite institutions, so that even within a broad range the average quality of an institution’s faculty is not terribly informative about individuals’ achievements.

Faculty at liberal arts colleges are also expected to do research. In a study including 51 schools and 439 faculty members, Bodenhorn (2003) showed sharp differences among institutions, with a citation-based count (most recent ten years of citations in the WoS) ranking colleges generally according with perceptions of their quality. While the statistics are difficult to compare, my best estimate is that eight full professors in these colleges received more citations than the median in Hamermesh’s (1989) sample of full professors in six large public research universities’ economics departments.

3.5 Differences among Journals

The history of citations to articles in J5 is suggested by Fourcade et al. (2015, Figure 4), showing that the AER and QJE have increased in relative importance since the early 1980s, while the JPE and Econometrica have decreased in relative importance (measured as shares of total citations to publications in the journal). This history, while very useful, does not consider differences across journals in the number of issues, articles per issue and length of articles. To account for these differences in the samples of articles in J5 in 1974-75 and 2007-08, I calculate each journal’s citations per AER2008-equivalent page, thus focusing not on the overall contribution of each journal, but instead on the attention paid to each published paper adjusted for its length. The resulting statistics might be viewed as article-based long-term impact factors.34

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33See Laband (2013) for a discussion and references on rankings of journals in economics.

34Looking at different points in the distributions of these statistics yields the same inferences.
Mean citations per AER2008-equivalent page to articles published in 1974-75 in the five journals are: AER, 7.0; Econometrica, 4.7; JPE, 9.4; QJE, 3.4; and REStud, 3.6. The mean citations per equivalent page in the 2007-08 sample are: AER, 2.3; Econometrica, 2.1; JPE, 2.0; QJE, 2.7; and REStud, 1.1. These calculations reinforce the inference that the QJE has increased in influence over this period, while the influence of articles published in the JPE and the REStud has decreased. Indeed, given that the REStud is such an outlier today, perhaps it makes sense for economists to refer to the “Top 4” journals and restrict the elite group in our discussions.

More important, however, we devote excessive attention, and excessive rewards, to publications in the J5. Inspired by Oswald (2007), consider articles published in the Economic Journal and the Review of Economics and Statistics in 2007-08, the same years as the later sample of J5, and using the same selection criteria (10+ pages, all in regular issues of journals, no comments, citations through calendar 2014). Obtaining the WoS citations to each of the 230 articles in this sample, I assigned each its percentile position in the distribution of J5 citations per AER2008-page equivalent. While the citations per page of the median-cited article in these journals ranks “only” at the 29th percentile of articles in the J5 (average citations per AER2008-equivalent page of 1.27), as Figure 7 shows there is substantial overlap with citations to J5 publications: Many articles in these two “lesser” general journals are cited more than the majority of articles published in J5, a result that holds for scholarly journals more generally (Stern, 2013).

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35Since the averages are based on citations through 2014, the citation counts for the earlier sample exceed those for the later sample. The opposite would be true if we only use the first seven years of citations to articles in the earlier sample, but the relative ranking of the journals in that earlier period would remain unchanged.


37As with the comparison to J5, there are substantial differences between these two journals. The median article in the EJ would be at the 23rd percentile of articles in J5, while the median article in the REStat would be at the 38th percentile. But one-fourth of the EJ articles are cited more per page than the median REStat article.
Just as there is substantial overlap in quality across economics faculties, so too there is substantial overlap in scholarly influence across journals. The main reason that a few economics journals are ranked more highly than all others is that a very few papers in these journals generate immensely more citations than other papers published in those journals or elsewhere. A very few outliers determine our perceptions of journal quality, perceptions that ignore the tremendous heterogeneity of articles within and across journals. (See also Verma, 2015.)

The WoS publishes “journal impact factors,” measuring for each journal in Year t the average citations per article published in Years t-1 and t-2 (although five-year impact factors, counting citations to articles published in Years t-1 through t-5, are also presented). Note first that impact factors ignore the heterogeneity in citations to individual articles across journals demonstrated above. Also, as Palacios-Huerta and Volij (2004) point out, ranking systems based on impact factors lack any basis in economic theory and are just one of very many ways of ranking journals’ influence.

The very short rear-view of the two-year calculation of impact factors severely disadvantages articles in economics compared to the “hard” sciences, given the extremely long lags between acceptance of an economics article and its publication. Indeed, even comparing within the social sciences this short rear-view and economists’ stinginess with citations disadvantages economics: The AER’s two-year impact factor was 3.3 in 2013, while those of the American Political Science Review and the American Journal of Sociology were 3.8 and 4.3 respectively. As broad indicators within a discipline, journal impact factors are useful, but they are of little use comparing across disciplines. Most important, they ignore the heterogeneity of the influence of articles published in the same journal.

4. Citations, Outcomes and Behavior

Citation analysis has been used in various ways to study outcomes within the economics profession and to examine more general issues that economists have focused on. Here I consider only the former: How citations—professional reputation—affect economists’ earnings and other rewards; how they have been used to pinpoint the importance of various scholarly contributions of major figures in the history of economic thought; and how subsequent citations have been used to examine the validity of decisions by
editors and professional committees about what to publish and whom to reward. I do not review the substantial literature in economics on innovation and patent citations, based on citations to research in the natural sciences (see, e.g., Trajtenberg, 1990); the sparse literature using citation analysis to understand the life cycle of scientists’ productivity (see, e.g., Levin and Stephan, 1991), nor the still sparser research program on the economics of legal institutions (Posner, 2000).

4.1 Salaries

A large literature has arisen examining the determinants of economists’ salaries, both in terms of the usual human capital inputs and direct measures of production (articles and books). Early examples of this genre are Johnson and Stafford (1974), Tuckman and Leahey (1975) and Hansen et al. (1978). My purpose in this sub-section is not to summarize this literature; instead, I focus on the much smaller part that has examined how citations affect salaries and whether their effects are independent of the role of numbers and quality of publications per se.

Examining the role of citations in determining academic salaries is interesting for several reasons. First, since the purpose of scholarship is to generate knowledge, and generated knowledge is reflected in citations, examining the impact of citations on salaries provides information on whether monetary incentives are in line with what we implicitly believe should be rewarded. Second, while other criteria should and will help determine salaries, a demonstration that citations as a measure of productivity are an important independent determinant might reassure scholars that objective reflections of their intellectual activities matter and that some notion of equity prevails in salary determination.

Estimating equations describing wage determination has been the stock-in-trade of labor economists for nearly a half century, with the basic equation including human capital measures—experience and educational attainment. All the studies discussed here included quadratics in (post-Ph.D.) experience, and all ignored educational attainment, since nearly all sample members have Ph.D. degrees. In all cases the dependent variable was the logarithm of nine-month salary.

There is a certain probably unavoidable homogeneity and narrowness in the salary-citations literature. Most important, because information on individuals’ remuneration in private universities and
colleges is not publicly available, all of the studies are based on economics faculties in public institutions; and only one is based on non-U.S. data. At the very least, results cannot be extrapolated to salary-setting outside the public sector. Moreover, the institutions studied are typically among the larger and more prestigious state schools, so that the results should not be extrapolated to salary determination in less well-known public institutions. Also, and no doubt because of the difficulties in obtaining data in the pre-internet age, only the most recent studies have fairly large samples of institutions and faculty members.

All studies attempt to measure the impacts of citations only on university salaries, usually only academic-year salaries. (Twelve-month salaries of “low-level” administrators were typically de-annualized for estimation, and “higher-level” administrators were typically excluded.) Thus outside professional income—e.g., from research grants, consulting, or book royalties—is not included (although Chan et al., 2014b, show that in academe generally the more heavily-cited receive higher non-academic earnings). To the extent that those who are more heavily cited obtain more than proportionate additional income from these sources, which seems likely, the estimated impacts of citations on salaries may understate their impacts on total professional income.

All of the studies summarized here used WoS (typically only SSCI) citation counts (since GS counts were not available at the time when most were conducted). Given the comparisons in Section 2, the reliance on the WoS instead of GS probably causes little bias. In both, however, as pointed out above, the difficulty in counting citations to frequently-appearing surnames may produce problems. Including these scholars in a study could introduce the errors into the estimated impacts of citations, while excluding them, as a few studies did, may have generated biased estimates of the nonlinear returns to citations if their citations depart from the average.

Within the WoS in these studies researchers had to choose the length of the individuals’ citation history to include in the estimation. In more recent research the choice was almost surely dictated by the much easier accessibility of (professional) lifetime than of annual citation counts. The opposite was true for pre-internet studies, when counting citations in annual hard-copy compendia made obtaining lifetime
counts much more difficult. Thus it is not surprising that earlier studies used a few recent years of citation counts, while more recent studies used lifetime totals.

Table 9 presents in chronological order a summary of the twelve studies of U.S. institutions and one of Ontario universities that included citations as determinants of economists’ salaries. Each holds constant a set of covariates that are listed in the table, and for each study I indicate in the last column its novelty in the literature. Most important, in the penultimate column in the table I summarize the impact of a ten-percent increase in the citation measure on the average sample member’s academic-year salary. The most important result answers the question whether citations matter in salary determination. Even in the first study of citations and salaries in economics (Hamermesh et al., 1982, which built upon Holtmann and Bayer, 1970, for the physical sciences), citations had a substantial positive impact on salaries at the same post-Ph.D. experience. A two-standard deviation in annual citations in that sample implied a nearly forty percent difference in salaries. While the estimates obviously differ across studies, within those differences they seem remarkably consistent, suggesting unequivocally that, through whatever mechanism, additional citations do increase salary.

Do citations increase salaries themselves, or is it simply that publication in “higher-quality” journals is rewarded and is positively correlated with citations? Some of the studies in Table 9, beginning with Sauer (1988), included covariates indicating the number, length and journal level of each scholar’s publications. Even in those cases, as the Moore et al. (1998), Baser and Pema (2003), Bratsberg et al. (2010), Hilmer et al. (2015), Gibson et al. (2015) studies, and the Sen et al. (2010) data, also suggest, the evidence generally suggests that citations have an independent impact on salaries.

Are all citations created equal, or does being cited in higher-quality journals (presumably by articles that themselves receive more scholarly attention) have a greater impact on salary? Kenny and Studley (1995) adjusted each citation in their counts by the impact factor of the journal where the citing article appeared. While they did not report effects of raw citation counts on salaries, the fact that the implied effect of their adjusted citation measure differed little from others shown in the table suggests that the locus of the
citation is unimportant. Moreover, a search over various values of discount rates implied that no discounting of citations exists (at least in salary-setting). Taken together, the results suggest that, in terms of impacts on salaries, a citation is a citation. The answer to the initial question of this paragraph is that citations alone independently affect salary, but much of their effect is through the average quality of the journal where an article appears. Nor does having citations concentrated on a few “home runs” have an extra impact on salary (Hamermesh and Pfann, 2012).

Books are not the major vehicle of scholarly communication in economics, and citations to them have been decreasing relative to journal articles (Rauber and Ursprung, 2008); but citations to them are included in GS totals. No study of economists’ salaries has separated out citations to books from those to articles, so the question of their role in salary-setting has not yet been answered. Indeed, the evidence on whether the publication of books itself even increases the academic salaries of economists is unclear, although some fairly convincing evidence (Ragan et al., 1999) for a small sample of public universities suggests that additional books and articles both increase salaries.

Does it matter when one’s work is cited? Siow (1991) showed that the impact on lifetime salary is greater for citations received earlier in one’s career. Given the evidence in Section 3 that most publications either “make a hit” or disappear from view, this suggests that earlier publication generates greater lifetime income. Whether this unsurprising result arises from the attention that early-publishing scholars achieve relative to their equally successful colleagues who publish later in their careers, or because academic salaries seem to be capped by within-university social norms, is unclear. Do incremental citations lead to incremental increases in salary, or is it merely that better scholars generally produce more citations and receive higher salaries? Hamermesh (1989) used longitudinal data to estimate models with person fixed effects and showed that much of the cross-section impact of citations on salaries remains.

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38 A similar adjustment by Bodenhorn (2003) hardly altered the citation-based ranking of economics departments in liberal-arts colleges.

39 A book was the most-cited work of two of the five most-cited economists in the sample of 1043 faculty members. Jin and Choi (2014) provide other evidence of the importance of books.
One might think that the market allocates talent so that the quality of scholars is lexicographic in school quality, and that the top schools are richer and pay higher salaries generally. (Of course, this implicit assumption about the allocation of academic talent is sharply contradicted by the evidence on x* in Table 8.) Several of the tabulated studies include institution fixed effects, so that any estimated impact on salaries was within-faculty. The estimates clearly show that, even within an economics department, being cited more raises one’s salary relative to otherwise observationally identical colleagues, other things equal.

Currently the citation count of each author of a co-authored article is increased when their article is cited. This practice and the apparent impact of citations on salaries leads to the question of whether citations affect salaries regardless of the number of authors of the cited article. The first article to examine this issue (Sauer, 1988) suggested that the returns to citations to a two-authored article were almost exactly half of those to citations to a solo-authored paper. (This result is consistent with the finding in Section 3 that in the 1970s the marginal productivity of an additional author was about zero.) Hilmer et al. (2015) examined the same issue, with results suggesting that citation counts are not discounted for co-authorship. This finding suggests market behavior that is inconsistent with the evidence in Section 3 on the marginal productivity of an additional author of articles published in 2007-08.

This literature raises the question why citation rates have independent (of the quantity and quality of publications) effects on salary. Most schools (Klein and Chiang, 2004, Hamermesh and Pfann, 2012) do not regularly obtain citation information for annual merit reviews, and Liebowitz (2014) reported on a similar survey of forty-six department chairs showing that citations are considered less important in decisions about promotion to full professor than are the outlets where the candidates’ articles appeared. One explanation is that, in a profession with substantial threatened and actual mobility, citations contribute directly to reputation which in turn enhances salaries through implicit market pressure.

4.2 Other Market Outcomes

The various studies of the impact of citations on salaries were made commensurable using an elasticity-like measure. Other effects of citations are too diverse to allow direct numerical comparisons. The nearest to the salary studies are the several pieces of research that have examined how citations affect
the likelihood of promotion, usually in the context of measuring gender differentials. Thus Ginther and Kahn (2004), using a sample of junior economists, showed that additional WoS citations, holding article quantity and quality constant, do increase the probability of having a tenured academic position ten years post-Ph.D., but the impact is small and statistically insignificant. Using GS data, de Paola and Scoppa (2015) showed that selections by boards of examiners for appointments as associate or full professor in Italy are very sensitive to the candidates’ h-statistics.

The age of citations might be viewed as an indicator of the rate of technical progress in a sub-specialty (or, taking a dimmer view, as showing the rate at which new fads appear). McDowell et al. (2011) used longitudinal data on the age of citations across eight sub-fields to examine individuals’ choices about whether to become a department chair and about activities post-chair position. Economists working in sub-fields in which citation age indicates more rapid depreciation of knowledge are less likely to become department heads; but if they do, they exit the position more slowly and are more likely to become higher-level administrators. Similar apparent rationality is shown by economists’ choices about how to treat refereeing requests. Those whose work is more heavily cited take longer to referee and are more likely to refuse editors’ refereeing requests (Hamermesh, 1995). Not only do additional citations affect salary; these results suggest that economists behave as if citations proxy the value of their time.

A substantial literature has examined whether scholarly achievement, as indicated by citations, affects the likelihood that an individual receives some academic honor or award. Effects on rewards may be especially important, given limits on monetary returns in academe and the visibility of awards and honors. There are two groups of decision-makers who determine these outcomes: Small committees (of oligarchs) that select individuals for awards (such as Nobel Prizes; AEA Distinguished Fellows, Clark Medalists and Presidents; and their European and other foreign equivalents) and electorates that explicitly choose among small sets of pre-selected candidates (e.g., for AEA vice-president or Executive Committee membership, or Econometric Society Fellow). The questions here are whether the choices of each type of decision-maker are at least partly based on achievement, measured by citations, and whether the importance of citations in these choices generally differs between the two types.
One might argue that scholarly achievement alone should determine choices of committees making selections for candidates in elections where voters determine outcomes or directly for awards/positions. The evidence does make it clear that scholarly achievement, as reflected by citations to the awardees’ work, is important in these choices. Thus Diamond and Toth (2007) showed that those AEA Presidents who have been selected from among earlier Executive Committee members have been more heavily (WoS) cited, conditional on the quantity of publications, age, location and Ph.D. “pedigree,” than others. Similarly, Hamermesh and Pfann (2012) demonstrated that among all full professors at 88 economics departments, those who are awarded Nobel Prizes, named AEA Distinguished Fellow or President, or Clark Medalist, have been more heavily (WoS) cited, conditional on a vector of their other characteristics. Given the sample size this result is unsurprising; but conditional on the total citation count, having more heavily concentrated citations—more “home runs”—also increases the likelihood of selection.

Relative citation rates also partly determine outcomes of elections in which association members or prior electoral winners vote for awards. Thus in a regional economics association in which voters chose within pairs of candidates, individual voters tended to choose candidates with higher relative citation rates (although various affinities, particularly gender among female voters, mitigated or even vitiated the importance of relative citations, as Dillingham et al., 1994, showed). Similar results, and a similar importance of gender, is shown by outcomes of the four-person elections for AEA Vice-Presidents and Executive Committee members (Donald and Hamermesh, 2006). In elections of Econometric Society Fellows, the more heavily-cited nominees from among lists that typically included around fifty economists were more likely to obtain the requisite thirty percent approval votes (Hamermesh and Schmidt, 2003).

Scholarly achievement, as measured by others’ citations to one’s publications, is a major determinant of awards, honors and elections in the economics profession, regardless of whether these outcomes are determined by committees or electorates. Candidates’ other characteristics are also important determinants, and no doubt one’s marketing efforts and the parochial preferences of committees or voters matter too. People can differ about the appropriate relative importance of scholarly achievement compared to these other factors, but the explicit recognition of one’s work by other scholars matters in these contexts.
4.3 Citations in the History of Economic Thought

Citation analysis has been used to evaluate the contributions of economists who are viewed today as historical figures, with the general aim being to infer both the magnitudes and the time paths of their influences on subsequent economic research. I organize this literature by whether the focus is on the total *oeuvre* of the historical figure or on some economic concept or construct that may have more than one origin, dealing with each in chronological order. Some of these studies are based on citation counts from one of the sources discussed here, but many are based on direct counts by the author from samples of journals (as in Stigler and Friedland, 1975). The former approach misses attributions that are not formally in footnotes or reference lists (e.g., the use of the term Marshallian), and even the latter approach catches these only if the historian of economic thought reads through all the potentially citing articles. Before discussing specific individuals and ideas, however, it is worth noting that Weinberg and Galenson (2005), using citations to economics Nobel Prize winners, showed that more empirically oriented scholars among this highly elite group did their most important work much later in life than those who created new, more abstract concepts.

Wight (2002) looked at journal articles’ references to Adam Smith in the last third of the 20th century, including directly counting as citations the use of terms like “Smithian” and “Smith’s” if the work/ideas cited were clearly those of Adam Smith. The fraction of journal articles referring to Smith’s work showed a steady increase through the end of the 20th century. No similar trend was observed by direct counts of SSCI citations, but, unlike citations to Malthus or Ricardo during this period, those to Smith did not decrease as a fraction of total SSCI citations. Clearly, there was something of a continuing Smith renascence during this period, perhaps stimulated by the 1976 bicentennial of the *Wealth of Nations*.

Biddle (1996) sampled articles in major journals from 1908-1950 to consider the influence of Wesley Clair Mitchell, most well-known today for his economic and statistical work on business cycles. Citations to Mitchell peaked during the 1930s (due to scholarly interest in the Great Depression?) but had tapered off by 1950. *Sic transit gloria mundi*. Using WoS citation counts Diamond (2004) examined the influence of Zvi Griliches’ research on subsequent scholars of technology, innovation and growth. As with
most economists’ work, citations to Griliches’ research increased initially, then fell off, except for the classic paper on hybrid corn (Griliches, 1957), which showed a nearly monotonic increase in citations through the early 2000s. This is one of the extremely rare papers for which the gloria did not seem to transit.

Duarte (2009) examined articles in major journals and in a few books to consider the role of Ramsey’s 1928 growth model on subsequent economic thought. After being essentially ignored for 20 years, Ramsay’s work on economic growth attracted large and rapidly growing numbers of citations beginning in the late 1950s. Interest peaked in the early 1970s with the boom in formal models of growth, then tapered off rapidly in the late 1970s as focus shifted away from those models. Stigler (1978) examined the use of the concept of the kinked oligopoly demand curve in subsequent literature. This idea, which surfaced in the late 1930s in the work of several economists, quickly commanded the attention of many other scholars, reaching a peak in the boom in neoclassical economics in the 1960s. By the 1970s, however, interest in it had diminished, probably as the use of game-theoretic approaches to understand non-competitive behavior became widespread.

Hoover (2014) searched through published articles and books to estimate the influence of Trygve Haavelmo’s “probability approach” to econometrics that was laid out in several articles in the early 1940s. As with most important contributions, its widespread and growing acceptance in its first post-publication decade was following by a decline in interest. In the decade before Haavelmo’s 1989 Nobel Prize, however, there was a resurgence of attention to the work, perhaps because of the growing controversy over “structural estimation,” a resurgence that had not tapered off much by the early 2000s. Diamond (1988) employed the SSCI to examine citations to Gerard Debreu’s work, particularly that on general equilibrium. The interesting focus was not on the time path of citations to this idea, but rather on the type of research that referenced it: Almost none of its influence was on subsequent empirical research—the research that cited it remained essentially isolated within the theory community. Finally, Claveau and Gingras (2017) used WoS citations to articles published from the 1950s onward to create what are essentially optimal partitions of the economic literature by sub-field, demonstrating the longevity of many sub-fields and the rise and fall of a few others, e.g., growth theory.
This is a fairly sparse literature; but it is one with substantial scope for additional work—the number of historical figures in economics is obviously increasing! With the growth and increasing ease of use of databases of citations, and, one might hope, with no further decline in economists’ focus on the origin of the ideas that underlie current research, citation analyses of economic-historical figures and ideas might accelerate.

4.4 Judging Editorial and Committee Decisions by Subsequent Citations

Since citations are a measure of quality, subsequent citations can provide a useful check on decisions about publishing, presenting awards, promotions and other outcomes. Many journals give awards (plaques or small honoraria) to the “best article” of the year. These awards are typically chosen by editorial or association committees. Are they given to the article with the most impact, or even to articles with impacts above the median? Coupé (2013) showed that award-winning papers are more likely than others to be the most WoS or GS-cited articles in a volume, but very few are the most-cited paper in a volume. Since the very fact that an article received an award may stimulate additional subsequent citations, this evidence suggests a somewhat tenuous relation to the citations that an article would have received absent the prize itself.

While we showed that several awards are partly determined by authors’ prior citations, one might wonder how well selection committees and voters succeed in choosing one or several winners among a set of similar candidates. Using Clark Medalists and newly-elected Econometric Society Fellows, Chan et al. (2014a) estimate double-differences between them and arguably comparable economists pre- and post-selection/election. The awardees/winners generate more and better subsequent citations than the control group, and even more citations per publication. Whether this double-difference reflects the perspicacity of committees and voters about awardees’/winners’ staying power, or simply that their “victory” per se leads others to cite their works more frequently, is unknowable from this statistical design.

The very acceptance of an article is itself an “award,” especially given the low acceptance rates even at second-level economics journals. Are these awards conferred based on successful perceptions of quality, or do they reflect, as so many economists believe (especially when their paper is rejected!) editorial
favoritism—toward friends and colleagues? Laband and Piette (1994) show that five-year WoS citations are greater, conditional on authors’ prior citations, to papers in top economics journals when the author and editor have some kind of connection—as student or colleague. Similar results are obtained by Colussi (2015), using eight years of post-publication WoS citations, and by Brogaard et al. (2014) using a much larger sample, defining “connection” as being colleagues with an editor, and accounting for school and even author fixed effects.

Editors’ choices also do not seem to be biased toward accepting papers that cite the editors’ own articles more. While it is true that scholars receive more citations in those journals and at those times when they are editing, a keyword count (Kim and Koh, 2014) suggests that these fillips to their citation counts do not arise from their choosing papers that cite them a lot. Perhaps instead authors select their submissions based on editorial interests, which guarantees more citations to the editor during his/her tenure.

There is no question that an article’s place in a journal issue matters: Lead articles receive many more citations than do subsequent articles (Smart and Waldfogel, 1996). But is this because their visibility leads readers to cite them more, or because editors’ decisions to place them in the lead reflects editorial beliefs about their quality, which are then validated in the market for research? Using a journal that ordered papers alphabetically (by author’s surname) in some issues, by editorial choice in others, Coupé et al. (2010) showed that about two-thirds of a lead article’s extra citations were due to its position, one-third to its inherent higher quality than subsequently placed articles in the same issue.

How we write our papers matters for their impact on others’ research. Taking a truncated sample of the top 1000 citation- and paper-ranked economists in the RePEc database, Olney (2017) shows, controlling for many individual differences, that those who are native English speakers are ranked higher. Even the title that one chooses for an article affects its subsequent (six-year, WoS) citations, suggesting that authors and editors should pay attention to the first thing that readers see in a journal article. Taking the three leading U.S. general journals, Mialon (2015) classifies articles as having a “poetic” title, one that includes some poetic device or figure of speech, as opposed to a purely informative title. The former are more heavily cited in the first six post-publication years, conditional on their authors’ prior citations, but
only among empirical papers. Finally, articles that draw from a wider literature of not directly networked research, perhaps a measure of innovativeness, are themselves more heavily cited subsequently (Boppart and Staub, 2016).

5. Main Points and Implications

Citations are just one possible measure of academic productivity; but they are a quantifiable and essentially market-based way of distinguishing among different participants in a particular academic enterprise. Obtaining accurate citations counts by whatever method—be it the WoS, GS profiles, RePEc or counting each person’s most-cited articles, requires substantial effort to ensure accuracy. The gratifying result here is that the particular citation count or statistic chosen yields a ranking that is very similar to other counts or rankings. Thus, for example, there is a very tight relationship between citation counts using the WoS and GS sources. So long as great care is taken in measuring people’s productivity by counting citations, individuals and institutions can be ranked appropriately and similarly using a number of different metrics. So too, any of these carefully constructed measures can be used to examine the relationship between this proxy for academic influence and outcomes in academe.

The overarching result throughout much of this article was the heterogeneity of citation counts across units being studied. There is substantial overlap in citations of articles that are published in journals that are viewed as being of different quality. Many of the articles in top journals receive little attention (though on average they receive more attention than those in lower-ranked journals). So too, many economists at lower-ranked faculties within what is already an elite group are cited more than the median faculty members at higher-ranked schools (though on average the former are cited less frequently than the latter). Indeed, these overlaps are the best justifications for examining citations—they allow for studying the achievements of individuals and their research instead of relying on opinions about average reputations of journals and institutions.

Fortunately the market for economists implicitly recognizes the importance of citations, since they help determine economists’ pay independently of the quantity and perceived quality of the articles that economists publish. But, while citations do matter in the market, do they matter enough, especially in salary-
setting, promotions and appointments? I would argue that they do not, and that the profession puts too much weight on publications in a few top journals. At a time when acceptance rates at the “Top Five” journals are around five percent, relying on editors’ parochial preferences in choosing which articles to publish when we assign credit in this profession may be inferior to letting the market for ideas, as reflected in citations, bear more weight in assigning elite monetary and reputational rewards. Indeed, that citations appear to matter more than the mere quantity or even the journal quality of publications in determining non-monetary rewards suggests that in certain circumstances this reflection of the market attains a greater importance.

No doubt the importance of average journal quality in salary and other decisions is due to the greater ease of obtaining information about it than about citations. What could be done to enhance the use of citations as measures of academic impact, to account for the obvious heterogeneity across articles, individuals and institutions? Asking faculty members to count and report citations annually is both demeaning and a creator of tedium (and unlikely to be successful). But asking faculty to register and create a GS profile gives them a simple discrete task and allows administrators to obtain and, I hope, use annual information on scholarly impact in determining salaries of current faculty and in targeting faculty elsewhere as potential recruits.

I am not arguing that citations should be the only determinant of rewards in economics. Indeed, a greater reliance on citations will lead some academics to increase their efforts to game the system to raise their citation counts. Rather, I am merely arguing that their impact is currently undervalued, a situation that is easily remediable. We remain at a point where the provision and use of more information on measured impacts of scholarly work would be better than less.

Economists have conducted a lot of research on the sociology of the economics profession, and research on how to count citations, citation-based rankings and citations as determinants of outcomes has burgeoned. Particularly interesting would be studies paralleling those summarized in Table 9, but concentrating on salary determination in private institutions. Is the impact of citations greater or less there than what has been found for public schools? Do citations affect one’s ability to obtain grant money or other non-nine-month professional income? In other words, do the percentage payoffs summarized in Table...
9 under- or overstate the returns to scholarly impact generally? The growth of web-based citation counts has made answers to these and other questions related to citations more readily obtainable.

Perhaps this judgment is too harsh, but very little of the research on the role of citations in economics has been used to shed light on more general economic issues. Just as the economics of particular sports has been used to study general issues in the functioning of labor and product markets, there is a lot of research to be done in which the empirical study of our citation practices can lead to greater understanding of such issues. Indeed, the general analysis of such diverse topics as the role of norms of interpersonal behavior, how discourse alters market behavior, and others, might be fruitfully studied using citation analyses as examples. In short, there is room for broad-ranging and broadly applicable research using citations in this profession.
REFERENCES


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Results found: 150

Sum of the Times Cited [?]: 2922
Sum of Times Cited without self-citations [?]: 2770

Citing Articles [?]: 2395
Citing Articles without self-citations [?]: 2321

Average Citations per Item [?]: 19.48

h-index [?]: 28

Figure 1. Screen from a May 2016 Search in the Web of Science
https://scholar.google.com/

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**Figure 2. Screen from a May 2016 Search in Google Scholar**
Figure 5. Calculation of an H-Statistic
WoSCitations = 7.860 + 0.182 GSCitations, Adj. $R^2 = 0.856$, N = 486
(0.930) (0.0034)

Figure 4. Relation between WoS Citations and Google Scholar Citations, Articles in J5 Journals, 2007-08*

*Excludes 11 articles with 1000+ GS citations.
WoSCitations = -0.318 + 0.173 GSCitations, Adj. $R^2 = 0.909$, N = 102
(2.81)   (0.0055)

**Figure 5. Relation between WoS Citations and Google Scholar Citations, Faculty in “Top Schools,” Ph.D. 2006 or Later***

*Excludes 7 faculty with 2000+ GS citations.
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Table 2. Distribution of Citation Measures of Faculty in 30 “Top” Economics Departments

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Table 3. Experience Profiles of Citation Measures in 30 “Top” Economics Departments

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<th>C5 (N = 1043)</th>
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<td>10th percentile</td>
</tr>
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<td>36-40</td>
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<td>41-45</td>
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<tr>
<td>46+</td>
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Table 4. Women and Citations, Means and Regression Estimates*

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<th>Experience&lt;18</th>
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<tr>
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<td>-352.5</td>
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<tr>
<td></td>
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<td>(577)</td>
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<td></td>
<td>-9.0</td>
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<tr>
<td></td>
<td>(66.6)</td>
<td>(72.6)</td>
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<tr>
<td>Rank of Dept.</td>
<td>-166.5</td>
<td>-34.3</td>
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<tr>
<td></td>
<td>(18.4)</td>
<td>(4.9)</td>
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<td>Pseudo- R²</td>
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<td>0.108</td>
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<tr>
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<tr>
<td></td>
<td>522</td>
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*These LAD regressions include a large vector of indicators of post-Ph.D. experience.
Figure 6a. Distribution of Citations per AER2008-Equivalent Page, J5 Articles, 1974-75*

Figure 6b. Distribution of Citations per AER2008-Equivalent Page, J5 Articles, 2007-08

*Excludes the top 4 percent of papers, with citations/equivalent page greater than 25
Table 5. “Resurrections” and “Flashes in the Pan” of J5 Articles 1974-75, N=444

<table>
<thead>
<tr>
<th>Rank in First 10 Years</th>
<th>Mean</th>
<th>Maximum</th>
<th>90th Percentile</th>
<th>Median</th>
<th>10th Percentile</th>
<th>Minimum</th>
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<tbody>
<tr>
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<td>372</td>
<td>244</td>
<td>295</td>
<td>372</td>
<td>424</td>
<td>426</td>
</tr>
<tr>
<td>Bottom Decile</td>
<td>366</td>
<td>235</td>
<td>288</td>
<td>366</td>
<td>426</td>
<td>441</td>
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<tr>
<td>9th Decile</td>
<td>339</td>
<td>90</td>
<td>217</td>
<td>363</td>
<td>431</td>
<td>442</td>
</tr>
<tr>
<td>2nd Decile</td>
<td>115</td>
<td>9</td>
<td>28</td>
<td>80</td>
<td>234</td>
<td>389</td>
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<tr>
<td>Top Decile</td>
<td>46</td>
<td>1</td>
<td>4</td>
<td>31</td>
<td>109</td>
<td>187</td>
</tr>
<tr>
<td>Top Ventile</td>
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<td>1</td>
<td>2</td>
<td>18</td>
<td>84</td>
<td>117</td>
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Table 6. Citations through 2014 by Number of Authors, J5 Articles, 1974-75 and 2007-08*

<table>
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<tr>
<th>N Authors</th>
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<th>3+</th>
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<td><strong>1974-75</strong></td>
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<td></td>
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<tr>
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<td>78.2</td>
<td>72.0</td>
<td>38.4</td>
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<tr>
<td>Std. Error</td>
<td>(12.2)</td>
<td>(9.9)</td>
<td>(23.5)</td>
</tr>
<tr>
<td>Median</td>
<td>20</td>
<td>38</td>
<td>10</td>
</tr>
<tr>
<td>90(^{th}) Percentile</td>
<td>171</td>
<td>164</td>
<td>88</td>
</tr>
<tr>
<td>Range</td>
<td>[0,2466]</td>
<td>[0, 679]</td>
<td>[0, 435]</td>
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<tr>
<td>N</td>
<td>312</td>
<td>114</td>
<td>18</td>
</tr>
<tr>
<td><strong>2007-08</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Mean</td>
<td>41.2</td>
<td>48.5</td>
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<tr>
<td>Std. Error</td>
<td>(3.2)</td>
<td>(3.4)</td>
<td>(6.3)</td>
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<tr>
<td>Median</td>
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</tr>
<tr>
<td>90(^{th}) Percentile</td>
<td>86</td>
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<td>128</td>
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<tr>
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<td>[2, 478]</td>
<td>[0, 365]</td>
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<tr>
<td>N</td>
<td>134</td>
<td>246</td>
<td>94</td>
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*Excludes papers from 1974-75 that were fewer than 5 nominal pages long, fewer than 10 nominal pages long in 2007-08.
Table 7. SSCI Citations per Article by Type in J5 Journals, 1974-75 and 2007-08*

1974-75 Publications, First 7 Years*

<table>
<thead>
<tr>
<th>Type</th>
<th>Theory</th>
<th>Empirical</th>
<th>Econometric Theory</th>
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</thead>
<tbody>
<tr>
<td>Mean</td>
<td>12.8</td>
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<tr>
<td>Std. Error</td>
<td>(1.3)</td>
<td>(1.9)</td>
<td>(1.3)</td>
</tr>
<tr>
<td>Median</td>
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<td>6</td>
</tr>
<tr>
<td>90th Percentile</td>
<td>29</td>
<td>38</td>
<td>25</td>
</tr>
<tr>
<td>Range</td>
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<td>[0, 109]</td>
<td>[0, 39]</td>
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<tr>
<td>N</td>
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<td>46</td>
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</table>

2007-08 Publications, First 7 Years

<table>
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<th>Theory w/ Simulation</th>
<th>Empirical: Borrowed Data</th>
<th>Empirical: Own Data</th>
<th>Experimental</th>
<th>Econometric Theory</th>
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</thead>
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<tr>
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<tr>
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<td>(9.3)</td>
<td>(5.7)</td>
<td>(5.4)</td>
<td>(7.9)</td>
<td>(6.8)</td>
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<tr>
<td>Median</td>
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<td>47</td>
<td>40</td>
<td>50</td>
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<tr>
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1974-75 Publications, Citations through 2014*

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<td>(8.9)</td>
</tr>
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<td>[0, 259]</td>
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<tr>
<td>N</td>
<td>280</td>
<td>113</td>
<td>46</td>
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</tbody>
</table>

*Excludes 5 articles that were not classifiable under any of these rubrics.
Table 8. Rankings and Overlaps in Faculty Quality across the Top 30 Economics Departments, Based on GS Citations to Each Author’s Five Most-Cited Publications (C5), and Ranks Using 50th or 90th Centiles

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<th>r_{d50}</th>
<th>C_{550}</th>
<th>KMS</th>
<th>x*</th>
<th>r_{d90}</th>
<th>r^*_{d50}</th>
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<td>30</td>
<td>267</td>
<td>30</td>
<td>12</td>
<td>30</td>
<td>26</td>
</tr>
</tbody>
</table>

*N is the number of faculty members included; r_{d50} is the department’s rank as measured by the median unadjusted citations, and C_{550} is that median; KMS (2003) is the ranking in the U.S. based on the impact factors of five years of publications in economics journals of all faculty at the institution; x* is the number of ranks above r_{d50} in which the department’s top quartile of faculty members, as ranked by citations, would be above the r_{d50} - x* department’s median faculty member; r_{d90} is the department’s rank as measured by the of the faculty member at the 90th percentile of this measure in the department; r^*_{d50} is the department’s rank as measured by the median experience-adjusted citations.
Figure 7. Position of Citations to Articles in the Economic Journal and Review of Economics and Statistics in the Distribution of Citations to J5 Articles
Table 9. The Impact of Citations on Economists' Salaries\textsuperscript{a}

<table>
<thead>
<tr>
<th>Study</th>
<th>Group and year(s)</th>
<th>Citation measure</th>
<th>Covariates</th>
<th>% impact of 10% increase in citations\textsuperscript{b}</th>
<th>Novelty</th>
</tr>
</thead>
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<tr>
<td>Hamermesh et al. (1982)</td>
<td>Fulls, 7 large public schools 1979</td>
<td>Average SSCI last 5 years</td>
<td>Experience, admin, theorist school FE (# articles, books)</td>
<td>1.4</td>
<td></td>
</tr>
<tr>
<td>Diamond (1986)</td>
<td>All ranks, 1961-79, UC-Berkeley</td>
<td>SSCI per year</td>
<td>Experience, cohort</td>
<td>2.6% for each extra 10 cites</td>
<td>Long longit.</td>
</tr>
<tr>
<td>Sauer (1988)</td>
<td>Assoc &amp; fulls, 7 large departments</td>
<td>SSCI last 7 years</td>
<td>Experience, admin, top journal, pages, other papers, books, school FE</td>
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<td></td>
</tr>
<tr>
<td>Hamermesh (1989)\textsuperscript{c}</td>
<td>Fulls, 6 large public schools 1979, 1985</td>
<td>Average SSCI last 5 years</td>
<td>Experience, admin, person FE</td>
<td>0.5</td>
<td>Person FE</td>
</tr>
<tr>
<td>Siow (1991)</td>
<td>Hamermesh (1989) data</td>
<td>Average SSCI last 5 years</td>
<td>Experience, admin, person FE</td>
<td>↓ as experience</td>
<td>Timing of citations in career</td>
</tr>
<tr>
<td>Kenny and Studley (1995)</td>
<td>All faculty, 10 large departments, 1987</td>
<td>SSCI lifetime</td>
<td>Experience, admin, gender, quality-adjusted articles</td>
<td>1.2</td>
<td>Impact-factor adjustment of citations</td>
</tr>
<tr>
<td>Author(s)</td>
<td>Type of Faculty</td>
<td>Schools/Fields</td>
<td>Database</td>
<td>Variables Included</td>
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<td>-------------------</td>
<td>----------------</td>
<td>----------------</td>
<td>----------</td>
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</tr>
<tr>
<td>Moore et al. (1998)</td>
<td>Tenured faculty, 9 state schools, 1992</td>
<td>SSCI lifetime</td>
<td>Experience, gender, seniority, quality/quantity, foreign, Ph.D. quality, teaching Award</td>
<td>0.4</td>
<td>Productivity and wage-seniority relation</td>
</tr>
<tr>
<td>Baser and Pema (2003)</td>
<td>All faculty, 9 midwestern state schools, 1992-1999</td>
<td>SSCI Lifetime</td>
<td>Gender, article quality/quantity, admin, school FE</td>
<td>0.8</td>
<td>Separate self, others’ citations</td>
</tr>
<tr>
<td>Bratsberg et al. (2010)</td>
<td>All faculty, 5 midwestern state schools, 1974-2004</td>
<td>SSCI Lifetime</td>
<td>Experience, gender, seniority, quality/quantity, pages, books</td>
<td>0.1</td>
<td>Wage-seniority and turnover</td>
</tr>
<tr>
<td>Hamermesh and Pfann (2012)</td>
<td>Fulls, 43 large public schools 2008</td>
<td>SSCI Lifetime</td>
<td>Experience, gender, foreign, field, no. of pubs, “home runs,” runs”, surname, name, school FE</td>
<td>0.9</td>
<td>Distribution of citations</td>
</tr>
<tr>
<td>Hilmer et al. (2015)</td>
<td>All faculty, 53 large public schools,</td>
<td>SSCI Lifetime</td>
<td>Experience, gender, seniority, quality/quantity, pages</td>
<td>0.8</td>
<td>Large sample, detail on co-authorship</td>
</tr>
<tr>
<td>Gibson et al. (2015)</td>
<td>Tenured, UC system</td>
<td>WoS Lifetime</td>
<td>Experience, gender, seniority, chaired, others</td>
<td>0.1</td>
<td>Journal impact factors</td>
</tr>
<tr>
<td>Based on Sen et al. (2010)</td>
<td>16 Ontario schools, a≥12</td>
<td>WoS Lifetime by Article</td>
<td>Experience, gender. 0.9 pooled 2006-2011 school FE</td>
<td>0.9</td>
<td>Canada</td>
</tr>
</tbody>
</table>

Degree of freedom is 9.

* log(9-month or 9-month equivalent salary) is the dependent variable.

*Unless otherwise noted.

*Uses log(9-month real compensation) as the dependent variable.