

Economic Research Evolves: Fields and Styles[†]

By JOSHUA ANGRIST, PIERRE AZOULAY, GLENN ELLISON, RYAN HILL,
AND SUSAN FENG LU*

Economic research has become more empirical, a shift documented by Hamermesh (2013), among others. We examine this shift in detail here, showing that it consists of changing research styles mostly within, rather than across, fields of economics. We also gauge the extent to which the shift in publication style is paralleled by a change in impact. Our analysis exploits a machine-learning-based classification of economics journal content into fields and styles, developed as part of a project analyzing citations to economics from other disciplines (Angrist et al. 2017).

The ability to classify papers automatically lets us take a broad look at economic research: our dataset includes 134,892 papers published in 80 journals between 1980 and 2015.¹ We present unweighted analyses of the full journal sample and weighted analyses that emphasize highly-cited journals.

* Angrist: Department of Economics, MIT, 50 Memorial Drive, Cambridge, MA 02142, and NBER (e-mail: angrist@mit.edu); Azoulay: Sloan School of Management, 100 Main Street, Cambridge, MA 02142, and NBER (e-mail: pazoulay@mit.edu); Ellison: Department of Economics, MIT, 50 Memorial Drive, Cambridge, MA 02142, and NBER (e-mail: gellison@mit.edu); Hill: Department of Economics, MIT, 50 Memorial Drive, Cambridge, MA 02142 (e-mail: ryanhill@mit.edu); Lu: Purdue University, 403 W. State Street, West Lafayette, IN 47907, and Northwestern University (e-mail: lu428@purdue.edu). We thank our extraordinarily efficient and skilled research assistants, Gina Li and Suhas Vijaykumar. Thanks also go to the AEA's Liz Braunstein for help with *Econlit*, to Ben Jones, Satyam Mukherjee, Brian Uzzi, and Heidi Williams, for help with the Web of Science data, and to our discussant, John Rust, for comments.

[†] Go to <https://doi.org/10.1257/aer.p.20171117> to visit the article page for additional materials and author disclosure statement(s).

¹ Journals in this set were among the 33 most cited by the *American Economic Review* in any of 1968, 1978, 1988, 1998, or 2008, plus other economics journals that were comparably well cited by the flagship journal of another discipline (such as the *American Political Science Review*).

I. Classification

Our analysis of publication and citation rates uses data from the Web of Science and Econlit databases pertaining to the articles published in our economics journal list.

A. Fields

Our field classification scheme exploits four sources of information: *JEL* codes; titles and keywords; the publishing journal; and the fields of the papers that a paper cites. We begin by compiling a training dataset that includes a set of papers for which Ellison's (2002) *JEL*-to-17-field mapping seems likely to be reliable. The training data also include papers in a few "field journals" (e.g., the *Journal of Labor Economics*) assigned to the field journal's field. A machine-learning algorithm trained on this dataset is used to generate an initial field classification for each paper. A clustering algorithm then uses data on each paper's initial classification and the initial classifications of the papers it cites to assign a final field (the online Appendix details all procedures used in this study).

The clustering algorithm is instructed to produce ten fields. Inspecting their contents, we label these fields microeconomics, macroeconomics, econometrics, public finance, labor, international, finance, industrial organization (IO), development, and miscellaneous. The miscellaneous field is an amalgam of several smaller fields including economic history, environmental economics, experimental economics, law and economics, political economy, and urban economics.

B. Styles

We classify papers as belonging to one of three research *styles*: theoretical, empirical, or econometrics. Papers in fields other than econometrics are classified as theoretical or empirical. We aim to label papers as "empirical" if they use

data to estimate economically meaningful parameters. Papers that cover methodological issues while also producing substantively meaningful estimates were also classified as empirical. To distinguish economic theory from econometric theory, papers classified in the econometrics field (using the process described above) were classified as falling into a distinct econometrics style.

Papers are also classified into styles by a machine-learning algorithm. Our training dataset for this purpose contains 5,850 papers: 1,507 hand-classified for use in Ellison (2002); and 3,343 additional randomly selected papers hand-classified mostly by our research assistants (who were also trained). We used this dataset to train a random forest algorithm that takes as input article titles, journal identifiers, the assigned field, *JEL* codes, keywords, the publication decade, and abstracts (where available).²

Our style classifications are less accurate in the 1970s, so we focus on post-1980 papers when reporting the distribution of publications by style and on post-1990 citing articles when reporting citations to styles. (Citations are backward-looking.)

C. Journal Weights

Different journals publish and cite different types of papers. Scholars are especially interested in the content of prestigious, highly-cited general interest journals, paying less attention to other journals. We use a weighting scheme to capture this hierarchy of journal importance; these “importance weights” are denoted w_j^t . For example, our importance-weighted measure of citations to style r is $c_r^t \equiv \sum_j w_j^t c_{jr}^t$, where c_{jr}^t is the fraction of journal j ’s year t citations which are to papers of style r .

The distribution of publications across fields and styles is reported using both weighted and unweighted counts. Like our weighted citation measure, the weighting scheme for publications produces a measure of output that emphasizes papers that appear in top journals. For example, the weighted measure of publications in field f is $s_f^t \equiv \sum_j w_j^t s_{jf}^t$, where s_{jf}^t is the fraction of journal j ’s publications which are in field f .

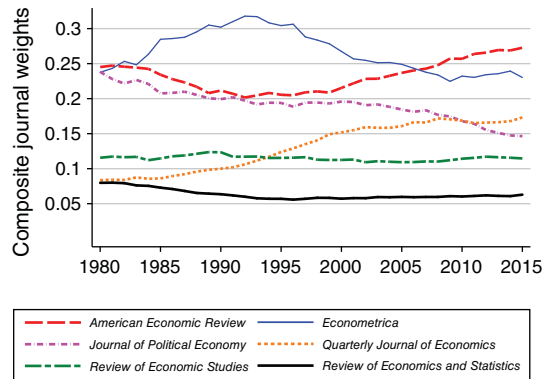


FIGURE 1. PAGE RANK WEIGHTING SCHEME

Note: The figure plots five-year moving averages of composite weights (μ_k^t) for each of the top six economics journals.

The importance weights, w_j^t , reflect the extent to which journal j (on our journal list) is cited by a weighted composite of six top journals. This imaginary “composite top journal” allows for changing importance within the top-six group (the top six includes the usual top five plus the *Review of Economics and Statistics*, which once rivaled the top five in importance). The first step in the construction of importance weights (detailed in the online Appendix) applies Google’s Page Rank algorithm to the matrix giving the fraction of each of the top six journal’s citations to the others. This produces a value, μ_k^t , for each of the six. We then compute a weighted average of citation fractions, $\tilde{w}_j^t \equiv \sum_k \mu_k^t c_{kj}^t$. The final w_j^t are the five-year moving averages of this series.

Figure 1 plots the smoothed time series of composite weights (μ_k^t) for each of the top six. Rising from rough parity with the *AER*, *Econometrica* leads during the late 1980s, peaking as the most important journal in the early 1990s, with declining weight through 2008. The *AER* and the *Quarterly Journal of Economics* grow in importance from about 1990 on. These trends—for example, the rise and fall of *Econometrica*—should be kept in mind when interpreting weighted results.

II. Economic Research Evolves

Descriptive statistics in the online Appendix show that fields interact with styles. Papers in the microeconomics field are mostly (though not entirely) classified as theoretical, while papers in

² Abstracts are unavailable prior to 1986.

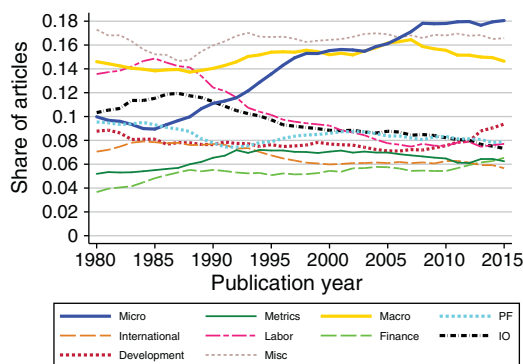


FIGURE 2. PUBLICATION SHARES BY FIELD

Note: The figure shows five-year moving averages of unweighted field shares.

the applied micro fields of labor, development, and public finance are mostly empirical. The collection of smaller fields grouped under the miscellaneous heading are nearly two-thirds empirical.

Figure 2 traces the evolution of economics journal output by field, showing the unweighted fraction of papers in each field among those published in the journals on our list between 1980–2015. Perhaps surprisingly, this figure shows the microeconomics field growing strongly since the mid 1980s, to the point where micro is now the largest field, bypassing macroeconomics in the mid-2000s. Microeconomics' increasing publication share reflects both a proliferation of theory journals and their increasing length.³ In contrast, the publication shares for labor and IO have both declined since the mid-late 1980s.

Which fields have the more influential journals been publishing? Figure 3 plots weighted field shares. Microeconomics has the largest share throughout, while macro also maintains a high share. The largest “applied micro” fields, labor, IO, and public finance, have declining weighted shares in the early years and no recent growth. In contrast with Figure 2, however, the importance-weighted statistics show substantial growth in our “miscellaneous” field. This amalgamated category includes environmental economics, experimental economics, urban eco-

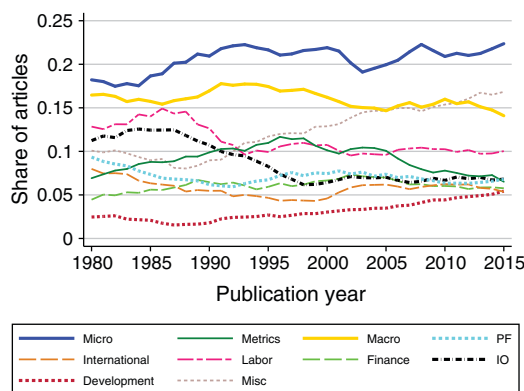


FIGURE 3. WEIGHTED PUBLICATION SHARES BY FIELD

Note: The figure shows five-year moving averages of weighted publication shares.

nomics, and political economy, fields that were once perhaps on the sidelines.

The empirical shift in economic scholarship is a *within-field* phenomenon, a pattern documented in Figure 4, which plots the weighted proportion of publications in each field classified as empirical. In the early 1980s, development and labor were the only fields in which the majority of weighted publications were empirical. The weighted empirical share has since grown in all fields, now exceeding 90 percent in labor and development. International and public finance are also now majority empirical. Even macroeconomics, criticized in the wake of the Great Recession for an excess of ivory-tower theorizing, has seen its empirical share grow by over 50 percent. In most fields, these trends reflect both increasing numbers of empirical papers and the improved journal placement of empirical work (excepting IO, whose unweighted empirical share is flat).

Changes in the overall empirical share also reflect within-field more than cross-field trends. Figure 5 plots unweighted publication style shares. Within-field shifts are muted somewhat by strong growth in the mostly theoretical microeconomics field. Even so, the share of economics publications devoted to empirical work, which held steady at about 50 percent from 1980 to 1995, has since increased to a little over 60 percent. With the share devoted to econometrics essentially unchanged, this increase came out of the theoretical share.

The weighted distribution of publications by style in Figure 6 shows a more dramatic rise in empirical work. In the early-mid 1980s, the

³For example, *Games and Economic Behavior* started in 1989 and *Economic Theory* in 1991. These two published 145 and 73 papers in 2014. *Journal of Economic Theory* has grown from 64 papers in 1980 to 130 in 2014.

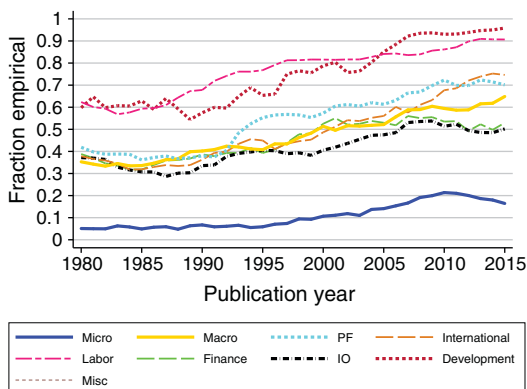


FIGURE 4. WEIGHTED FRACTION EMPIRICAL BY FIELD

Note: Five-year moving averages of the weighted fraction of publications in each field that are empirical.

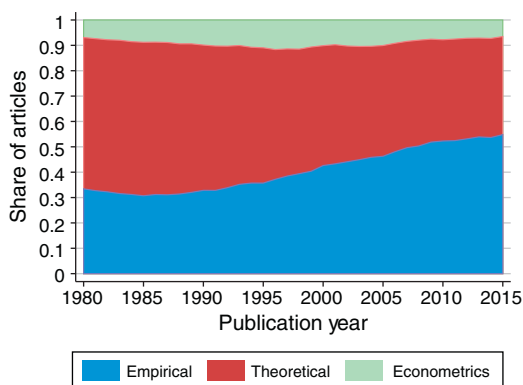


FIGURE 6. WEIGHTED PUBLICATIONS BY STYLE

Note: Five-year moving averages of weighted publication shares in each style.

weighted empirical share was only around one-third. This lower starting point reflects the fact that empirical papers were once disproportionately found in less-cited journals. The weighted empirical share has increased steadily since around 1985, and now exceeds 55 percent.

The shift in citation shares toward empirical work is even stronger than the publication shift. Figure 7, which reports the weighted distribution of styles of cited papers for citations made in 1990–2015, shows empirical work garnering less than 30 percent of weighted citations in 1990. (At the time, about half of recent publications were empirical.) The empirical citation share is now almost 50 percent. With roughly 10 percent of

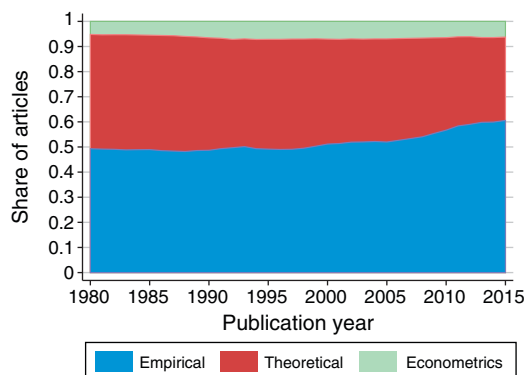


FIGURE 5. PUBLICATIONS BY STYLE

Note: Five-year moving averages of unweighted publication shares in each style.

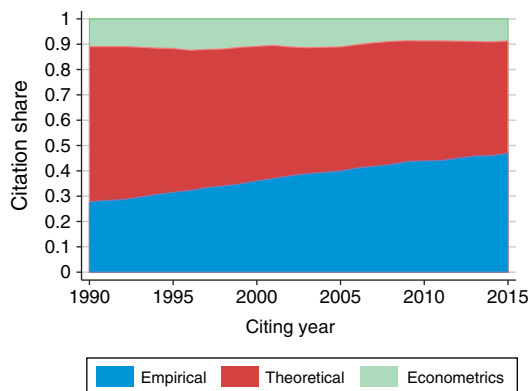


FIGURE 7. WEIGHTED CITATIONS BY STYLE

Note: Five-year moving averages of weighted citation shares in each style.

citations going to econometrics, empirical work is now cited more often than theoretical work. Increased citations to empirical work naturally reflect the fact that more cited papers are empirical. But this change also reflects movement of empirical work into better, more-cited journals.

III. Citations Per Paper

An analysis of citations per paper highlights the different dimensions of increasing empirical impact. For each paper i published in year $t(i)$, we model the conditional mean of weighted citations to this paper (c_i) as an exponential function of style dummies (Emp_i and Met_i), a vector X_i of

article-level covariates, and a battery of year-specific field and journal indicators, indexed by $f(i)$ and $j(i)$. Baseline controls include a cubic polynomial for article page length and dummies for the number of authors. The model is

$$E[c_i | X_i, Emp_i, Met_i, f(i), j(i), t(i)] \\ = \exp[\beta_1^t Emp_i + \beta_2^t Met_i + \beta_3^t X_i + \delta_{j(i)}^t + \gamma_{f(i)}^t].$$

Because many papers are never cited and the citation distribution is highly skewed, an exponential model fits the conditional mean function of interest better than a linear model (37 percent of the papers are never cited by other papers in the sample). The coefficient β_1^t captures a time-varying covariate-adjusted log ratio of empirical to theoretical citations per paper.

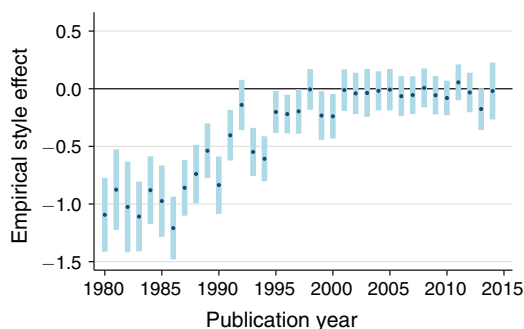
Theoretical work published in the 1980s and 1990s was cited far more often than empirical work of the same period. This can be seen in panel A of Figure 8, which plots the time series of estimates of β_1^t from a model omitting field and journal effects. Relative citation rates to empirical work grew steadily starting in the late 1980s, but only around the year 2000 did citation rates for empirical papers reach parity with citation rates for theoretical work.

The estimates of a model with field and journal controls reported in panel B of Figure 8 show that much of the theoretical citation advantage can be attributed to differences in the distribution of papers across fields and journals. Controlling for field and journal dummies—that is, looking within fields and journals—the empirical citation deficit shrinks to less than 50 percent in the early 1980s and disappears in the late 1980s. After 2000, empirical papers are cited more than theoretical work in the same field that was published in the same journal and year. The increasing attention to empirical work therefore reflects factors beyond improved journal placement or persistent field-specific citation norms.

IV. Summary

Using machine-learning methods to classify economics papers into fields and styles, we document major shifts in research output and the types of papers referenced. The growth in empirical work reflects a substantial shift within rather than across fields. Microeconomics remains the

Panel A. Controls for length and number of authors



Panel B. Including field and journal controls

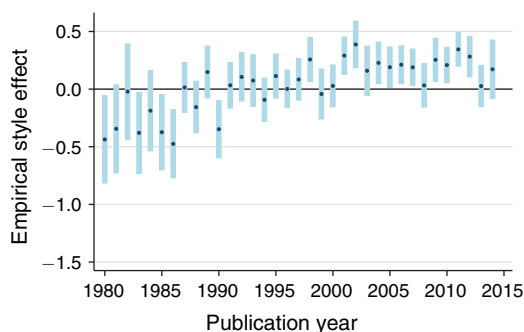


FIGURE 8. EMPIRICAL/THEORY CITATIONS PER PUBLICATION

Notes: Poisson regression estimates of the empirical effect on weighted citations per paper. Panel A estimates are from models estimated separately by year, with flexible controls for paper length and number of authors. Estimates in panel B add field and journal controls. Confidence bands use robust standard errors.

largest field, while some applied micro fields have shrunk. But more empirical papers are being published and they are appearing in more influential journals. Citations to empirical work have grown even more than empirical output, although the empirical share of citations is just now reaching 50 percent.

REFERENCES

- Angrist, Joshua, Pierre Azoulay, Glenn Ellison, and Susan Feng Lu. 2017. "Inside Job or Deep Impact? Assessing the Influence of Economics Scholarship." Unpublished.
- Ellison, Glenn. 2002. "The Slowdown of the Economics Publishing Process." *Journal of Political Economy* 110 (5): 947–93.
- Hamermesh, Daniel S. 2013. "Six Decades of Top Economics Publishing: Who and How?" *Journal of Economic Literature* 51 (1): 162–72.

This article has been cited by:

1. Quan-Hoang Vuong, Anh-Tuan Bui, Viet-Phuong La, Minh-Hoang Nguyen, Hung-Hiep Pham, Thanh-Hang Pham, Thi-Hanh Vu, Thu-Trang Vuong, Manh-Toan Ho. 2021. Mirror, mirror on the wall: is economics the fairest of them all? An investigation into the social sciences and humanities in Vietnam. *Research Evaluation* **30**:1, 57-72. [[Crossref](#)]
2. Matthias Aistleitner, Stephan Puehringer. 2021. The Trade (Policy) Discourse in Top Economics Journals. *New Political Economy* **26**:5, 748-764. [[Crossref](#)]
3. Syed Hasan, Robert Breunig. 2021. Article length and citation outcomes. *Scientometrics* **126**:9, 7583-7608. [[Crossref](#)]
4. Patrick Button, LaPorchia A. Collins, Augustine Denteh, Mónica García-Pérez, Ben Harrell, Elliott Isaac, Engy Ziedan. 2021. Teaching controversial and contemporary topics in economics using a jigsaw literature review activity. *The Journal of Economic Education* **91**, 1-10. [[Crossref](#)]
5. Jesus Crespo Cuaresma, Harald Oberhofer, Rupert Sausgruber. 2021. Special issue: Evidence based policy making—selected papers of the 2020 Annual Meeting of the Austrian Economic Association. *Empirica* **48**:3, 589-591. [[Crossref](#)]
6. Kilian Buehling. 2021. Changing research topic trends as an effect of publication rankings – The case of German economists and the Handelsblatt Ranking. *Journal of Informetrics* **15**:3, 101199. [[Crossref](#)]
7. Martin Paldam. 2021. Methods Used in Economic Research: An Empirical Study of Trends and Levels. *Economics* **15**:1, 28-42. [[Crossref](#)]
8. Maja Micevska. 2021. Revisiting forced migration: A machine learning perspective. *European Journal of Political Economy* **54**, 102044. [[Crossref](#)]
9. Jakob Kapeller, Stephan Puehringer, Christian Grimm. 2021. Paradigms and policies: the state of economics in the German-speaking countries. *Review of International Political Economy* **49**, 1-27. [[Crossref](#)]
10. Sascha Schweitzer, Jan Brendel. 2021. A burden of knowledge creation in academic research: evidence from publication data. *Industry and Innovation* **28**:3, 283-306. [[Crossref](#)]
11. Achim Ahrens, Christopher Aitken, Mark E. Schaffer. Using Machine Learning Methods to Support Causal Inference in Econometrics 23-52. [[Crossref](#)]
12. Xiaohui Wei, Wanling Chen, Xiao Li. 2021. Exploring the financial indicators to improve the pattern recognition of economic data based on machine learning. *Neural Computing and Applications* **33**:2, 723-737. [[Crossref](#)]
13. Paul Lagneau-Ymonet, Bénédicte Reynaud. 2020. The making of a category of economic understanding in Great Britain (1880–1931): ‘the unemployed’. *Cambridge Journal of Economics* **44**:6, 1181-1196. [[Crossref](#)]
14. Fuqiang Zhang, Xiaole Wu, Christopher S. Tang, Tianjun Feng, Yue Dai. 2020. Evolution of Operations Management Research: from Managing Flows to Building Capabilities. *Production and Operations Management* **29**:10, 2219-2229. [[Crossref](#)]
15. Guangsa Jin, Chenle Li, Ya Sun. 2020. Exploring the Macrostructure of Research Articles in Economics. *IEEE Transactions on Professional Communication* **63**:3, 227-243. [[Crossref](#)]
16. Gad Yair, Keith Goldstein. 2020. The Annus Mirabilis paper: years of peak productivity in scientific careers. *Scientometrics* **124**:2, 887-902. [[Crossref](#)]
17. Michael Thicke. 2020. Evaluating Formal Models of Science. *Journal for General Philosophy of Science* **51**:2, 315-335. [[Crossref](#)]
18. Georges Bresson. Comments on “An Econometrician’s Perspective on Big Data” by Cheng Hsiao 431-443. [[Crossref](#)]

19. Emily C. Marshall, Anthony Underwood. 2020. Is economics STEM? Trends in the discipline from 1997 to 2018. *The Journal of Economic Education* 51:2, 167-174. [[Crossref](#)]
20. Adam Jones, Peter Schuhmann, Daniel Soques, Allison Witman. 2020. So you want to go to graduate school? Factors that influence admissions to economics PhD programs. *The Journal of Economic Education* 51:2, 177-190. [[Crossref](#)]
21. Thiago Dumont Oliveira, Marwil J. Dávila-Fernández. 2020. From modelmania to datanomics? The rise of mathematical and quantitative methods in three top economics journals. *Scientometrics* 123:1, 51-70. [[Crossref](#)]
22. Josh Angrist, Pierre Azoulay, Glenn Ellison, Ryan Hill, Susan Feng Lu. 2020. Inside Job or Deep Impact? Extramural Citations and the Influence of Economic Scholarship. *Journal of Economic Literature* 58:1, 3-52. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]
23. Diego Mendez-Carbajo. 2020. Active learning with FRED data. *The Journal of Economic Education* 51:1, 87-94. [[Crossref](#)]
24. Daniel Levy, Tamir Mayer, Alon Raviv. 2020. Academic Scholarship in Light of the 2008 Financial Crisis: Textual Analysis of NBER Working Papers. *SSRN Electronic Journal* . [[Crossref](#)]
25. Lukas Bäuerle, Stephan Pühringer, Walter Otto Ötsch. Mathematik und Grundlagenveranstaltungen 107-129. [[Crossref](#)]
26. Olivér Kovács. 2019. Grounding Complexity Economics in Framing Modern Governance. *Acta Oeconomica* 69:4, 571-594. [[Crossref](#)]
27. Florentin Glötzl, Ernest Aigner. 2019. Six Dimensions of Concentration in Economics: Evidence from a Large-Scale Data Set. *Science in Context* 32:4, 381-410. [[Crossref](#)]
28. Ofer H. Azar. 2019. The influence of psychological game theory. *Journal of Economic Behavior & Organization* 167, 445-453. [[Crossref](#)]
29. Markus Seyfried, Florian Reith. Mixed Methods for Research into Higher Education: Solving the Problem of Institutionalized Introspection? 111-127. [[Crossref](#)]
30. Shaen Corbet, Michael Dowling, Xiangyun Gao, Shupe Huang, Brian Lucey, Samuel A. Vigne. 2019. An analysis of the intellectual structure of research on the financial economics of precious metals. *Resources Policy* 63, 101416. [[Crossref](#)]
31. Magda Fontana, Fabio Montobbio, Paolo Racca. 2019. TOPICS AND GEOGRAPHICAL DIFFUSION OF KNOWLEDGE IN TOP ECONOMIC JOURNALS. *Economic Inquiry* 57:4, 1771-1797. [[Crossref](#)]
32. John B. Davis. 2019. Economics and economic methodology in a core-periphery economic world. *Brazilian Journal of Political Economy* 39:3, 408-426. [[Crossref](#)]
33. Lutz Bornmann, Klaus Wohlrabe. 2019. Normalisation of citation impact in economics. *Scientometrics* 120:2, 841-884. [[Crossref](#)]
34. Luciana de Souza Leão, Gil Eyal. 2019. The rise of randomized controlled trials (RCTs) in international development in historical perspective. *Theory and Society* 48:3, 383-418. [[Crossref](#)]
35. Fabian Braesemann. 2019. How behavioural economics relates to psychology – some bibliographic evidence. *Journal of Economic Methodology* 26:2, 133-146. [[Crossref](#)]
36. Dennis A. V. Dittrich. 2019. Verhaltensökonomik als Gegenprogramm zur Standardökonomik?. *List Forum für Wirtschafts- und Finanzpolitik* 44:4, 841-859. [[Crossref](#)]
37. Jan Schnellenbach. 2019. Klassische Institutionenökonomik: Kommentar zum Beitrag von Wolfram Elsner. *List Forum für Wirtschafts- und Finanzpolitik* 44:4, 827-834. [[Crossref](#)]

38. Emily C. Marshall, Anthony Underwood. 2019. Writing in the discipline and reproducible methods: A process-oriented approach to teaching empirical undergraduate economics research. *The Journal of Economic Education* **50**:1, 17-32. [[Crossref](#)]
39. Alice Louise Kassens. Taking a Path Less Traveled: Mastering 'Metrics Without a Textbook 137-153. [[Crossref](#)]
40. Sven Langedijk, Ian Vollbracht, Paolo Paruolo. The Potential of Administrative Microdata for Better Policy-Making in Europe 333-346. [[Crossref](#)]
41. John B. Davis. 2019. Economics and Economic Methodology in a Core-Periphery Economic World. *SSRN Electronic Journal* . [[Crossref](#)]
42. Jo Thori Lind. How Do Economists Think? 269-281. [[Crossref](#)]
43. Fuqiang Zhang, Xiaole Wu, Christopher S. Tang, Tianjun Feng, Yue Dai. 2019. Evolution of Operations Management Research Topics and Methods in the Digital Economy. *SSRN Electronic Journal* . [[Crossref](#)]
44. Matthias Aistleitner, Stephan Puehringer. 2019. Exploring the Trade (Policy) Narratives in Top Economics Journals. *SSRN Electronic Journal* . [[Crossref](#)]
45. Fritz Schiltz, Chiara Masci, Tommaso Agasisti, Daniel Horn. 2018. Using regression tree ensembles to model interaction effects: a graphical approach. *Applied Economics* **50**:58, 6341-6354. [[Crossref](#)]
46. Ioana E. Marinescu, Patrick N. Lawlor, Konrad P. Kording. 2018. Quasi-experimental causality in neuroscience and behavioural research. *Nature Human Behaviour* **2**:12, 891-898. [[Crossref](#)]
47. José Edwards, Yann Giraud, Christophe Schinckus. 2018. A quantitative turn in the historiography of economics?. *Journal of Economic Methodology* **25**:4, 283-290. [[Crossref](#)]
48. Angela Ambrosino, Mario Cedrini, John B. Davis, Stefano Fiori, Marco Guerzoni, Massimiliano Nuccio. 2018. What topic modeling could reveal about the evolution of economics. *Journal of Economic Methodology* **25**:4, 329-348. [[Crossref](#)]
49. Beatrice Cherrier, Andrej Svorenčik. 2018. The quantitative turn in the history of economics: promises, perils and challenges. *Journal of Economic Methodology* **25**:4, 367-377. [[Crossref](#)]
50. François Claveau, Catherine Herfeld. 2018. Network Analysis in the History of Economics. *History of Political Economy* **50**:3, 597-603. [[Crossref](#)]
51. Scott M Swinton. 2018. Why Should I Believe Your Applied Economics?. *American Journal of Agricultural Economics* **100**:2, 381-391. [[Crossref](#)]
52. Ernest Aigner, Matthias Aistleitner, Florentin Gloetzel, Jakob Kapeller. 2018. The Focus of Academic Economics: Before and after the Crisis. *SSRN Electronic Journal* . [[Crossref](#)]
53. Ali Sina nder, Sergey V. Popov, Sascha Schweitzer. 2018. Leadership in Scholarship: A Machine Learning Based Investigation of Editors' Influence on Textual Structure. *SSRN Electronic Journal* . [[Crossref](#)]
54. Ali Sina nder, Sascha Schweitzer, Hakan Yilmazkuday. 2018. Specialization, Distance, and Research Quality in Collaborations: So Far Yet so Close. *SSRN Electronic Journal* . [[Crossref](#)]
55. Jan Brendel, Sascha Schweitzer. 2017. The Burden of Knowledge in Economics. *SSRN Electronic Journal* . [[Crossref](#)]