



**University of
Zurich**^{UZH}

**Zurich Open Repository and
Archive**

University of Zurich
University Library
Strickhofstrasse 39
CH-8057 Zurich
www.zora.uzh.ch

Year: 2020

Foreign competition and domestic innovation: evidence from US patents

Autor, David ; Dorn, David ; Hanson, Gordon H ; Pisano, Gary ; Shu, Pian

DOI: <https://doi.org/10.1257/aeri.20180481>

Posted at the Zurich Open Repository and Archive, University of Zurich

ZORA URL: <https://doi.org/10.5167/uzh-196248>

Journal Article

Accepted Version

Originally published at:

Autor, David; Dorn, David; Hanson, Gordon H; Pisano, Gary; Shu, Pian (2020). Foreign competition and domestic innovation: evidence from US patents. *American Economic Review: Insights*, 2(3):357-374.

DOI: <https://doi.org/10.1257/aeri.20180481>

Foreign Competition and Domestic Innovation: Evidence from US Patents[†]

By DAVID AUTOR, DAVID DORN, GORDON H. HANSON, GARY PISANO,
AND PIAN SHU*

Manufacturing accounts for more than three-quarters of US corporate patents. The competitive shock to this sector emanating from China's economic ascent could in theory either augment or stifle US innovation. Using three decades of US patents matched to corporate owners, we quantify how foreign competition affects domestic innovation. Rising import exposure intensifies competitive pressure, reducing sales, profitability, and R&D expenditure at US firms. Accounting for confounding sectoral patenting trends, we find that US patent production declines in sectors facing greater import competition. This adverse effect is larger among initially less profitable and less capital-intensive firms. (JEL F14, L11, L60, O19, O31, O34, P33)

Despite accounting for less than one-tenth of US private nonfarm employment, US manufacturing still generates more than two-thirds of US R&D spending and corporate patents.¹ In light of China's spectacular growth in manufacturing exports, the potential impact of import competition on innovation by domestic firms has gained relevance.² In this paper, we study the effect of rising import competition on US innovation at the firm level.

How product market competition affects innovation is of long-standing theoretical interest.³ Schumpeterian models posit that a more competitive product market reduces firm profit margins, resulting in lower investments in innovative activity (e.g., Dasgupta and Stiglitz 1980). But as Arrow (1962) noted, when competitive

* Autor: MIT and NBER (email: dautor@mit.edu); Dorn: University of Zurich and CEPR (email: david.dorn@econ.uzh.ch); Hanson: Harvard University and NBER (email: gordon_hanson@hks.harvard.edu); Pisano: Harvard University (email: gpisano@hbs.edu); Shu: Georgia Institute of Technology (email: pian.shu@scheller.gatech.edu). Pete Klenow was coeditor for this article. We are grateful to Rohan Thavarajah, Alex He, and Ante Malenica for excellent research assistance. Autor and Hanson acknowledge funding from the National Science Foundation (SES-1227334), and Autor also acknowledges funding from the Alfred P. Sloan Foundation (#2011-10-120). Dorn acknowledges funding from the Swiss National Science Foundation (BSSGI0-155804 and CRSII1-154446). Pisano and Shu acknowledge funding from the Division of Faculty Research and Development at Harvard Business School.

[†] Go to <https://doi.org/10.1257/aeri.20180481> to visit the article page for additional materials and author disclosure statement(s).

¹ Helper, Krueger, and Wial (2012) compute a manufacturing share in US R&D spending of 68 percent based on data from the National Science Foundation's Business R&D and Innovation Survey. In our data, manufacturing accounts for 71 percent of all corporate patents with US-based inventors and an application year of 2007.

² A substantial literature shows that import competition from China has contributed to job loss in US manufacturing (e.g., Bernard, Jensen, and Schott 2006; Autor, Dorn, and Hanson 2013; Autor et al. 2014; Pierce and Schott 2016; Acemoglu et al. 2016).

³ See Gilbert (2006) and Cohen (2010).

pressure is low and pre-innovation rents are high, firms may have little incentive to invest in innovation. Aghion et al. (2005) formalize these competing insights, showing how differences between pre- and post-innovation rents determine a firm's response to competition. This response depends on the dispersion in technological advancement across firms. When dispersion is low, intensified competition encourages firms to innovate to "escape competition." When dispersion is high, more competition may stifle innovation among laggard enterprises.⁴

Following a large literature (Cohen 2010), we measure innovation using firm patenting and R&D expenditure, which we supplement with firm-level data on sales, investment, and profitability.⁵ We match the assignees of all US patents granted between 1975 and 2013 to publicly held firms listed in Compustat through 2014 using a search-engine-based algorithm that disambiguates inconsistently spelled and abbreviated renditions of firm names on patent applications. Following Autor et al. (2014), we isolate the industry-specific component of US import growth that is driven by export-supply growth in China by instrumenting rising US import competition with contemporaneous changes in industry-level import penetration in other high-income countries. Using alternative identification strategies (Bloom, Draca, and Van Reenen 2016; Pierce and Schott 2016), we obtain comparable results.

We first show that increased imports from China ramped up competitive pressure on publicly listed US firms, reducing US and global sales, diminishing book and stock values, curtailing purchases of labor and capital inputs, and proportionally reducing R&D investment. We next show that this increase in competitive pressure led US firms to *decrease* their output of innovations as measured by patent grants. These negative effects are confirmed when using two alternative measures of trade exposure based on changes in trade policy, neither of which is likely to be driven by technology trends: the US grant of permanent normal trade relations (PNTR) to China in 2000 and the scheduled elimination of Multi-Fiber Agreement (MFA) quotas in 2005. Notably, the induced fall in patents is roughly proportional to the contraction of sales, employment, capital, and R&D expenditure at import-competing firms.

To overcome the fact that our firm-level analysis necessarily excludes patents granted to firms that are *not* listed in Compustat, we exploit the observation that all patents are assigned a technology class regardless of corporate ownership, which allows us to estimate the effect of import competition on patenting at the *technology class* level. Consistent with the firm-level results, we find that technology classes with greater growth in import penetration saw a substantial relative decline in *corporate* patents during our analytic window of 1991–2007. We find no similar decline in patenting by noncorporate entities (e.g., governments and universities), indicating that our key findings are specific to private sector innovation and do not reflect underlying correlations between innovation opportunities and rising trade exposure.

⁴The prediction is unclear for firms that are technological leaders. Additional causal mechanisms raise further ambiguity. See Hart (1983), Bloom et al. (2014), and Chen and Steinwender (2019) as well as Shu and Steinwender (2019) for a synthesis of this literature.

⁵On using patents to measure innovation, see Jaffe and Trajtenberg (2002) and Moser (2016). On the economic value of patents, see Hall, Jaffe, and Trajtenberg (2005) and Kogan et al. (2017).

Our results stand in contrast to those of Bloom, Draca, and Van Reenen (2016) for European firms and highlight the underlying theoretical ambiguity in the relationships between competition and innovation. We speculate that the differential impacts of foreign competition on domestic innovation in the United States versus Europe may be reconciled through the lens of Aghion et al. (2005): US industries have been found to display larger gaps in the technological capabilities of leading and lagging firms when compared to firms in Europe (Bartelsman, Haltiwanger, and Scarpetta 2013; Hashmi 2013). We provide suggestive evidence in support of the Aghion et al. (2005) model by showing that within the United States, the negative effects of import competition concentrate on firms that are initially less productive and less profitable.

Our work is contemporaneous to emerging evidence on the impacts of import competition on innovation-related outcomes in North America.⁶ Xu and Gong (2017) and Hombert and Matray (2018) also use Compustat data to show that import competition has a negative impact on firms' R&D expenditure and financial outcomes. Xu and Gong (2017) further show that R&D is reallocated toward more productive firms, and Hombert and Matray (2018) find that firms with large R&D stocks mitigate the negative impact of import competition through product differentiation. Kueng, Li, and Yang (2016) find that more trade-exposed Canadian firms experience strong declines in self-reported innovation outcomes. We contribute to this literature by providing a comprehensive analysis of the impact of Chinese import competition on both output (patenting) and input (R&D expenditure) of firm innovation in the United States. Our analysis accounts for industry pre-trends and establishes robustness to a wide range of identification strategies. Consistent with a competitive market response, we find that Chinese import competition reduces innovation at both the firm and technology class levels in the US private sector.

I. Measuring Trade Exposure

To measure import penetration, we match trade data to US manufacturing industries. In a second step, we match industry-level trade exposure to firm-level data, which we then match to patent records. The online Appendix provides further details.

To create measures of changing import penetration, we match trade data to four-digit Standard Industrial Classification (SIC) US manufacturing industries using the UN Comtrade Database and the crosswalk in Pierce and Schott (2012). Our baseline measure of trade exposure is the change in the import penetration ratio for a US manufacturing industry over the period 1991 to 2007, defined as

$$(1) \quad \Delta IP_{j\tau} = \frac{\Delta M_{j,\tau}^{UC}}{Y_{j,91} + M_{j,91} - E_{j,91}},$$

where for US industry j , $\Delta M_{j,\tau}^{UC}$ is the change in imports from China over two subperiods, 1991 to 1999 and 1999 to 2007, and $Y_{j,91} + M_{j,91} - E_{j,91}$ is initial

⁶An extended set of results are available in our earlier working paper, Autor et al. (2017).

absorption (industry shipments, $Y_{j,91}$, plus industry imports, $M_{j,91}$, minus industry exports, $E_{j,91}$) at the start of the period. We choose 1991 as the starting year for the analysis as it is the earliest period for which we have disaggregated bilateral trade data to match to US manufacturing industries.⁷

Observed changes in import penetration may in part reflect domestic shocks to US industries that determine both US import demand and innovative activity. Even if the dominant factors driving China's export growth are internal supply shocks, US industry import demand shocks may still contaminate bilateral trade flows. To capture this supply-driven component in US imports from China, we follow Autor et al. (2014) and instrument for trade exposure in (1) with the variable

$$(2) \quad \Delta IPO_{j\tau} = \frac{\Delta M_{j,\tau}^{OC}}{Y_{j,88} + M_{j,88} - X_{j,88}},$$

where $\Delta M_{j,\tau}^{OC}$ is the growth in imports from China in industry j during the period τ for a group of eight industrialized countries that does not include the United States.⁸ The denominator in (2) is initial absorption in the industry in 1988. The motivation for the instrument in (2) is that high-income economies are similarly exposed to growth in imports from China that is driven by supply shocks such as expanding product variety, falling prices, rising quality, and diminishing trade and tariff costs in China's surging sectors. The identifying assumption is that industry import demand shocks are uncorrelated across high-income economies. Autor, Dorn, and Hanson (2013) and Autor et al. (2014) provide evidence on the robustness of this instrumentation approach in studying the China trade shock. We complement this strategy by also using the identification approaches of Pierce and Schott (2016) and Bloom, Draca, and Van Reenen (2016) as discussed in Section IIIC.

II. Evidence of Increased Competitive Pressure from Imports

Prior to assessing the impact of rising Chinese import competition on *realized* innovation outcomes, we verify that this force exerts significant competitive pressure on import-competing firms, measured by sales, profitability, and R&D investment.⁹ For our analysis, we match industry-level trade exposure to Compustat North America, which provides industry affiliation and financial statement data on companies whose shares are traded at a North American stock exchange.¹⁰ We estimate the following regression model using these matched data:

⁷Our empirical approach requires trade data reported under Harmonized System (HS) product codes in order to match with US SIC industries. The year 1991 is the earliest in which many countries began using the HS classification.

⁸These countries are Australia, Denmark, Finland, Germany, Japan, New Zealand, Spain, and Switzerland, which represent the high-income countries for which we can obtain disaggregated bilateral HS trade data back to 1991.

⁹These results corroborate literature on the consequences of the China trade shock. See footnote 2, Xu and Gong (2017), and Hombert and Matray (2018).

¹⁰Compustat covers companies that have at least one regularly, actively, and publicly traded issue listed on a US or Canadian exchange and that provide a regular filing of financial reports. The data include foreign companies that use American Depository Receipts. The coverage of our data ends in 2014.

TABLE 1—EFFECT OF CHINESE IMPORT COMPETITION ON FIRM OUTCOMES, 1991–2007 AND 1975–1991 (FOR FALSIFICATION TEST). DEPENDENT VARIABLE: CHANGE IN SALES, INPUTS, FIRM VALUE, AND R&D (IN PERCENTAGE POINTS)

	Firm sales		Firm global inputs		Firm value		R&D
	United States (1)	Global (2)	Workers (3)	Capital (4)	Book (5)	Market (6)	Investment (7)
<i>Panel A. Exposure period 1991–2007</i>							
ΔUS industry exposure to Chinese imports	–1.29 (1.07)	–0.79 (0.41)	–0.75 (0.41)	–1.28 (0.55)	–1.42 (0.71)	–1.65 (0.76)	–0.95 (0.39)
Mean outcome variable	32.94	54.42	19.92	55.55	–14.80	41.38	50.77
Observations	2,200	3,098	2,803	3,104	2,965	2,679	2,279
<i>Panel B. Pre-period 1975–1991</i>							
ΔUS industry exposure to Chinese imports	0.47 (0.31)	0.48 (0.40)	0.21 (0.27)	0.80 (0.49)	0.54 (0.58)	0.28 (0.50)	0.79 (0.41)
Mean outcome variable	45.44	60.63	3.55	58.19	–4.84	58.73	72.87
Observations	1,971	2,188	2,084	2,178	2,162	1,872	1,453

Notes: Every regression in panel A comprises two stacked first differences, 1991–1999 and 1999–2007, and includes a period dummy and 11 indicator variables for broad sectors of manufacturing. The data includes all US firms for whom the indicated outcome variable is reported in Compustat at the start and end of a given period. The relative change of an outcome variable is defined as the first difference in the outcome over a period $t, t + 1$, divided by the average of the outcome across the two periods t and $t + 1$. Panel B provides falsification tests that regress the change in outcomes during the periods 1975–1983 and 1983–1991 on the future increase in Chinese import penetration, which is averaged over the 1991–1999 and 1999–2007 periods. All models are weighted by a firm's US-inventor patents, averaged over the start and end of a period. Standard errors are clustered on four-digit SIC industries.

$$(3) \quad \Delta Y_{ij\tau} = \alpha_{\tau} + \beta_1 \Delta IP_{j\tau} + \gamma X_{ij0} + e_{ij\tau},$$

where $\Delta Y_{ij\tau}$ is the change in a firm outcome for firm i in industry j over time period τ , defined as $100 \times (Y_{ij,t1} - Y_{ij,t0}) / (0.5Y_{ij,t1} + 0.5Y_{ij,t0})$, and $\Delta IP_{j\tau}$ is growth of import exposure (in percentage points) for industry j over period τ , as defined in equation (1) and instrumented by $\Delta IPO_{j\tau}$, as defined in equation (2).¹¹ The control vector X_{ij0} comprises time trends for 11 US manufacturing sectors. For consistency with the analysis of firms' patenting activity, observations are weighted by the number of firm patents, averaged over the start and end year of period τ , though we obtain similar results when observations are weighted by firms' US sales.

Table 1 reports results. The first two columns of panel A indicate that firms facing an exogenous rise in Chinese import competition exhibit a relative decline in both US and global sales between 1991 and 2007. The former effect is imprecisely measured, while the latter estimate implies that global sales contract by eight-tenths of a percent for each percentage point increase in Chinese import penetration. Columns 3 and 4 report a corresponding decline in firm utilization of labor and capital, suggesting an overall decrease in firms' scale of operation. Consistent with these contractions, the book and stock market value of exposed firms falls by approximately 1.5 percent for every percentage point rise in import competition (columns 5 and 6). Column 7 indicates that R&D investment at import-competing firms drops roughly one for one with rising import exposure, indicating a fall in *inputs* used in innovation.

¹¹ We use the proportional change scaling of the outcome variable rather than the log change because some values of Y are equal to zero.

Could these patterns merely reflect firm-level trends that predate the China shock? Panel B of Table 1 indicates otherwise. When we regress firm outcomes during the *pre-period* of 1975–1991 on *future* change in Chinese imports occurring between 1991 and 2007, we find that in seven of eight cases, the point estimates are insignificant and in all cases the coefficients are opposite signed to those in panel A.

III. The Effect of Industry-Level Import Competition on Firm-Level Patenting

A. Firm-Level Data and Patent Matching

We combine Compustat data with utility patents from the US Patent and Inventor Database, which covers all patents granted by the US Patent and Trademark Office between 1975 and March 2013.¹² A long-standing limitation of patent records is that the firm names entered on patents frequently contain unusual abbreviations or misspellings, which inhibit machine matching. We overcome this problem with an algorithm that leverages the fact that internet search engines function as repositories of information on common spelling variations of company names.¹³ If “International Bussiness Machines” is a common misspelling of IBM, an internet search engine will suggest IBM.com or IBM’s Wikipedia page as its top search results. Our algorithm enters the firm names appearing on patents and Compustat records into the Bing.com search engine and harvests the URLs of the search results. This procedure allows us to match patents to firms based on shared web search URLs in cases where the name strings on patents and firm records do not match exactly.¹⁴ Compared to the traditional string matching and manual inspection methods used in the NBER Patent Data Project, our procedure significantly improves efficiency without sacrificing accuracy. Since our search-engine-based algorithm minimizes the need for manual intervention, it is readily scalable. Our approach is generalizable to the matching between any two firm-level datasets and to many other applications in which string matching is complicated by spelling variations. Section A of the online Appendix discusses our matching procedure in detail.

Our analysis uses utility patents *applied* for in the years 1975, 1983, 1991, 1999, and 2007. The 1991–1999 and 1999–2007 periods coincide with the surge in Chinese import competition; the 1975–1983 and 1983–1991 periods enable us to account for industry pre-trends. Since the mean difference between the patent application and grant dates is 2.5 years in our data (standard deviation 1.5 years), right censoring due to not observing patents granted after March 2013 is unlikely to pose a serious problem. We use inventors’ addresses listed on patents to determine the location of invention and restrict our patent sample to corporate patents with a US-based primary inventor. We match 72 percent of these corporate patents to firms that appear in Compustat.¹⁵ Our patent sample consists of more than 170,000 patents

¹²The data are available at <https://github.com/funginstitute/downloads>. See Li et al. (2014) for a description.

¹³On supervised approaches to matching patents to disambiguating patent assignees, see also Ventura, Nugent, and Fuchs (2015); Morrison, Riccaboni, and Pammolli (2017); and Balsmeier et al. (2018).

¹⁴In the case of IBM, we identify 147 name variations on its patents (online Appendix, Table A1).

¹⁵Unmatched corporate patents were likely granted to firms that were never public and thus lack a Compustat record.

that originate at 6,081 firms, which jointly accounted for 95 percent of all R&D expenditure that Compustat recorded in 1991.

Panel A of Figure 1 plots by year of patent application the total number of patents with US-based primary inventors, corporate patents with US inventors, and corporate patents with US inventors matched to Compustat. All three series show a sharp rise between 1983 and 1999 and a modest decline between 1999 and 2007.¹⁶ The fourth series in the figure shows that the decline of patenting in the early 2000s coincides with a strong acceleration in US imports from China.

The aggregate trends in patenting shown in Figure 1, panel A mask important heterogeneity in the shifting concentration of patenting across sectors. Between 1975 and 2007, patenting in computers and electronics expanded rapidly while patenting in chemicals and petroleum was stagnant. As a result, computers overtook chemicals as the sector that produces most patents in the early 1990s (online Appendix, Table A4).¹⁷ Panel B of Figure 1 plots the change in log patents for 1991–2007 against the contemporaneous change in import penetration for computers and electronics, chemicals and petroleum, and all other manufacturing industries. Since the growth in Chinese imports was much stronger in computers than in chemicals, the raw correlation between patenting and trade exposure at the broad sectoral level is *positive*. However, panel B of Figure 1 also indicates a positive raw correlation between sectoral patent growth in the *pre-period* of 1975–1991 and the *subsequent* growth in Chinese import competition between 1991 and 2007. The acceleration of patenting in the computer sector and the stagnation in the chemical sector thus each predate the rise in Chinese imports. Accounting for these secular (pre-)trends will be critical to our identification.

B. Import Competition and Patent Production: Baseline Estimates

Following equation (3), we estimate the impact of rising Chinese import competition on patent production at the firm level. Panel A of Table 2 applies a bare-bones specification that includes no covariates beyond the change in import penetration and a year-specific constant.¹⁸ The conditional correlation between the change in firm patents and the change in industry import penetration is positive for 1991–1999 (column 1), negative for 1999–2007 (column 2), and negative for the stacked first-difference model that includes both periods with a separate time effect for each period (column 3). The difference between the OLS and IV estimates reported in rows (a) and (b) is modest.

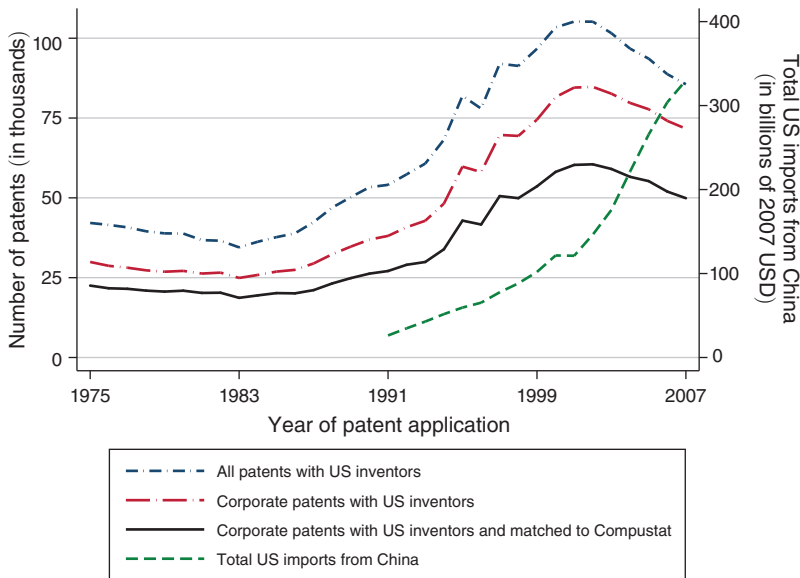
Panel B of Table 2 confronts the pre-trends in patenting in the computer and chemical sectors documented in Figure 1, panel B. Merely including indicator variables for these two trending sectors (computers and chemicals) reveals a robust negative relationship between changes in industry import penetration and changes in

¹⁶The literature offers several possible explanations for the slowdown in patenting in the early 2000s. See Jorgenson, Ho, and Stiroh (2008); Gordon (2012); Boldrin and Levine (2013); and Arora, Belenzon, and Pataconi (2018).

¹⁷On the growth of information technology and software patents, see Jorgenson (2001) and Bessen and Hunt (2007).

¹⁸The sample in Table 2 includes all firms that applied for at least one patent at the start or end of a given period. Continuous coverage in Compustat during the period is not required. Table A5 in the online Appendix alternatively considers a balanced panel of Compustat-covered firms that is also used for the analysis in Table 1.

Panel A. Number of patents by application year



Panel B. Growth in sectoral patenting and import penetration from China

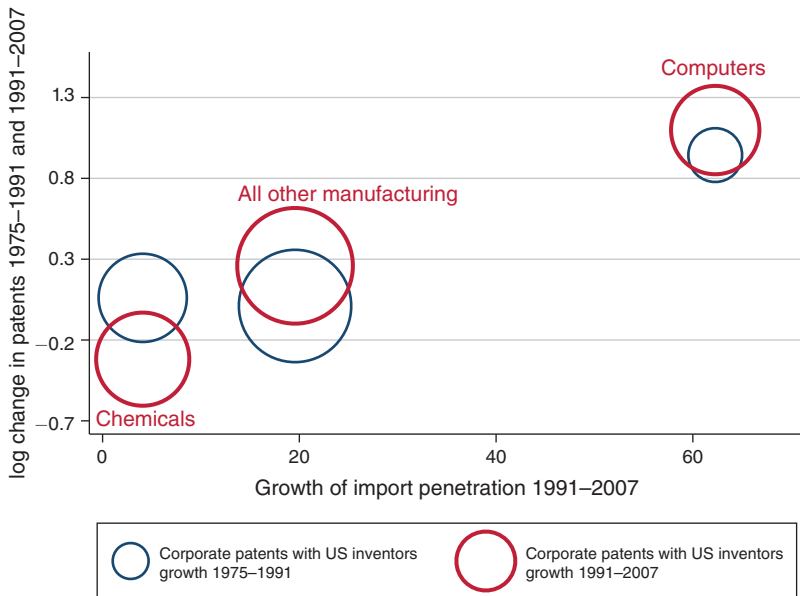


FIGURE 1. TRENDS IN PATENTING

firm patenting. Specifically, both OLS and IV models (rows c and d) detect a statistically significant negative effect of Chinese import competition on firm-level domestic patent production in both subperiods and the stacked first-difference specification. IV estimates are uniformly larger (more negative) than their OLS counterparts, suggesting that the IV models purge simultaneity or measurement error (or both).

TABLE 2—EFFECTS OF CHINESE IMPORT COMPETITION ON FIRM-LEVEL PATENTING, 1991–2007 AND 1975–1991 (FOR FALSIFICATION TEST): OLS AND INSTRUMENTAL VARIABLES MODELS. DEPENDENT VARIABLE: CHANGE IN CORPORATE PATENTS BY US-BASED INVENTORS (PERCENTAGE POINTS) RELATIVE TO MID-PERIOD NUMBER OF PATENTS

	Exposure period: 1991–2007			Pre-period: 1975–1991			$\Delta(1991-2007)-$ $(1975-91)$
	1991– 1999 (1)	1999– 2007 (2)	1991– 2007 (3)	1975– 1983 (4)	1983– 1991 (5)	1975– 1991 (6)	
<i>Panel A. Models without controls</i>							
(a) OLS, no controls	1.37 (1.14)	–0.45 (0.22)	–0.32 (0.26)	0.91 (0.33)	1.09 (0.61)	1.02 (0.45)	–1.34 (0.57)
(b) 2SLS, no controls	0.40 (1.39)	–0.29 (0.40)	–0.26 (0.40)	1.06 (0.43)	1.70 (0.68)	1.44 (0.54)	–1.70 (0.67)
<i>Panel B. Controlling for computer and chemical sector main effects</i>							
(c) OLS, 2 mfg sector dummies (computers, chemicals)	–0.87 (1.02)	–0.63 (0.12)	–0.91 (0.15)	0.36 (0.25)	–0.30 (0.64)	0.00 (0.38)	–0.92 (0.38)
(d) 2SLS, 2 mfg sector dummies (computers, chemicals)	–2.36 (1.40)	–0.57 (0.31)	–1.25 (0.53)	0.36 (0.33)	0.17 (0.61)	0.27 (0.40)	–1.53 (0.56)
<i>Panel C. Adding detailed controls</i>							
(e) 2SLS, 11 mfg sector dummies	–1.77 (1.16)	–0.46 (0.34)	–1.10 (0.51)	0.54 (0.41)	0.47 (0.68)	0.52 (0.49)	–1.62 (0.62)
(f) 2SLS, 11 mfg sector dummies + industry controls	–1.10 (1.26)	–0.50 (0.34)	–1.11 (0.48)	0.60 (0.44)	0.38 (0.55)	0.50 (0.43)	–1.61 (0.55)
(g) 2SLS, 11 mfg d. + industry/ firm controls	–1.16 (1.06)	–0.52 (0.34)	–1.17 (0.48)	0.62 (0.42)	0.33 (0.58)	0.48 (0.43)	–1.65 (0.54)
(h) 2SLS, 11 mfg d. + industry/ firm controls + technology mix	–1.13 (1.31)	–0.72 (0.35)	–1.35 (0.50)	0.31 (0.34)	0.27 (0.62)	0.27 (0.39)	–1.63 (0.55)
(i) 2SLS, 11 mfg d. + industry/ firm controls + technology mix + 2 lags	–1.29 (1.27)	–0.80 (0.39)	–1.39 (0.47)	n/a	n/a	n/a	n/a
Mean outcome variable	65.37	–7.61	24.42	–18.65	36.46	10.33	19.88
Observations	4,157	4,114	8,271	2,437	3,035	5,472	13,743

Notes: Each coefficient is derived from a separate firm-level regression of the relative change in patents on the change of Chinese import penetration. The relative change in patents is defined as the first difference in patents over a period t , $t + 1$, divided by the average number of patents across the two periods t and $t + 1$. Import penetration increased by a mean of 2.07 (standard deviation 4.37) percentage points in 1991–1999, 6.46 (standard deviation 14.34) in 1999–2007, and 4.54 (standard deviation 11.34) pooled over both periods. Columns 4–6 provide falsification tests that regress the change in patents on the future increase in Chinese import penetration averaged over the 1991–1999 and 1999–2007 periods. Columns 3 and 6 present stacked first-differences models for the periods 75–83/83–91 and 91–99/99–07 and include a period dummy, while column 7 indicates the difference between the import exposure coefficients of the column 3 and 6 models. Models (c) and (d) include dummies for the computer/communication and chemical/petroleum industries. Model (e) includes a full set of dummies for 11 manufacturing sectors. Model (f) additionally includes five industry-level controls for production characteristics (production workers as a share of total employment, log of average wage, and the ratio of capital to value added, all measured at the start of each period, as well as computer investment and investment in high-tech equipment, both expressed as a share of total investment and measured in 1990 for the models of columns 1–3 and in 1972 for the models of columns 4–6). Model (g) additionally includes a dummy variable for US-headquartered firms and controls for the log US sales of a firm and for its global R&D expenditure expressed as a share of global sales. It also includes two dummy variables indicating firms for which the two latter controls are not available in the Compustat data. Model (h) additionally controls for the fraction of a firm's patents that falls into each of the six major patent technology categories defined by Hall, Jaffe, and Trajtenberg (2001), averaged over start-of-period and end-of-period patents. Model (i) additionally controls for two eight-year lags of the outcome variable. All models are weighted by a firm's US-inventor patents averaged over the start and end of a period. Standard errors are clustered on four-digit SIC industries.

We document the importance of pre-trends in columns 4–6 of Table 2, where we regress changes in firm-level patenting in the *preceding* 16-year (pre-China shock) period of 1975 to 1991 (in two 8-year intervals) on the change in industry-level import penetration over the subsequent 1991–2007 period. In both 8-year time intervals as well as in the stacked model, and for both the OLS (row a) and 2SLS (row b) specifications, we find a positive, statistically significant relationship between changes in firm-level patenting and China import growth 16 years later. Notably, simply adding sector dummies for computers and chemicals reduces this correlation to near zero and renders it statistically insignificant across all columns (rows c and d). Column 7 shows that we can also eliminate the influence of these pre-trends by double differencing the post- and pre-China shock outcomes to estimate the effect of China import competition on the change in firm-level patenting in 1991–2007 relative to 1975–1991. This approach requires no sectoral-level dummies (rows a and b) and is little affected by their inclusion (rows c and d).

Panel C of Table 2 introduces additional controls that might mitigate the relationships detected in panel B, including (row e) dummy variables for 11 manufacturing sectors, (row f) measures of industry factor and technology intensity at the start of the period (share of production workers in industry employment, log capital over value added, log average industry wage, computer investment as a share of overall investment, and high-tech equipment as a share of total investment),¹⁹ (row g) firm start-of-period characteristics (a dummy for US-headquartered firms, log US sales, and firm global R&D spending as a share of firm global sales), (row h) the technology mix of a firm's patents (the fractions of a firm's patents that fall into each of six major technology fields), and (row i) the firm's prior patent growth using the 8-year and 16-year lags of the outcome variable.²⁰ Results from these specifications are comparable in magnitude to the parsimonious 2SLS model in row (d) that includes only sectoral dummies to absorb patenting trends in computers and chemicals.

C. Alternative Identification Strategies

We obtain causal identification by exploiting the external supply-driven component of rising Chinese exports to the United States, identified by the industry-level covariance between rising Chinese import penetration to the United States and other high-income countries. This interpretation would be threatened if an exhaustion of technological opportunities in specific industries occurring worldwide rendered these “mined out” industries vulnerable to Chinese competition. To address this concern, we implement Pierce and Schott's (2016) strategy of leveraging changes in trade policy uncertainty related to China's attainment of PNTR with the United States in 2000 to instrument for the acceleration of the growth in US industry import

¹⁹The first three variables are based on the NBER productivity database and the latter two on Feenstra and Hanson (1999).

²⁰Following Hall, Jaffe, and Trajtenberg (2001), we group patents into six technology fields based on their primary technology class: chemical, computers and communications, drugs and medical, electrical and electronics, mechanical, and others. To maintain a constant sample size across specifications, missing values for the firm or industry controls are replaced with a value of zero, and an indicator variable for each missing control is added to the regression.

competition in the 2000s versus the 1990s.²¹ As a second strategy, we exploit the phaseout of import quotas from the MFA between 1999 and 2005 as an additional source of within-industry changes in import exposure, following the approach of Bloom, Draca, and Van Reenen (2016). These MFA quotas applied primarily to imports from the textile, apparel, and leather sectors.

Table 3 reports side-by-side estimates of the impact of import competition on firm-level patenting, estimated first by OLS and then with each of our three instrumentation strategies. Column 1 presents an OLS regression of the change in firm-level patenting on the change in import penetration and the full vector of control variables from Table 2. This estimate detects a highly significant negative relationship between import competition and innovation that is comparable in magnitude to the corresponding 2SLS estimate in Table 2 (column 3, row h). Column 2 adds a full set of four-digit industry fixed effects that absorb any secular trend in patenting at the detailed industry level. The coefficient on the interaction term between import exposure and the 1999–2007 indicator variable in column 2 is now identified by *within-industry, over time* changes in import exposure and patenting. The significant negative point estimate of -0.71 means that industries that saw a greater increase in import penetration in 1999–2007 relative to 1991–1999 saw a larger fall in patent production in the latter period relative to the former.

Columns 3 and 4 apply our baseline China exposure instrument to this exercise.²² These estimates are negative and highly significant, as expected. This reduced form IV inference is also robust to the inclusion of four-digit industry main effects (column 4) so that identification comes from within-industry changes in import exposure during 1999–2007 versus 1991–1999.

We next implement the Pierce and Schott (2016) identification strategy by exploiting the gap between most favored nation (MFN) and non-MFN tariffs as a shock to industry-level China import competition facing US firms after 2000. Since the elimination of the non-MFN Smoot–Hawley Tariff threat improved the competitive position of Chinese exporters to the US market, this gap measure should have a negative impact on domestic US patenting following China's attainment of PNTR in 2000. Because we have no strong prior on the relationship between the tariff gap and US domestic innovation during the *pre-2000* period, we include both a main effect for the tariff gap instrument and an interaction of that instrument with the 1999–2007 dummy.²³ Column 5 shows that the tariff gap is positively correlated with industry-level patenting growth during the 1990s, prior to China's WTO accession. Relative to this trend, however, industries with higher tariff gaps experienced a statistically significant slowdown in patenting in the 2000s

²¹ Had China's MFN trading status not been reauthorized in any year after 1980, its US tariff rates would have risen to the far-higher Smoot–Hawley rates set in the 1930s. Pierce and Schott (2016) and Handley and Limão (2017) show that US industries with higher gaps between non-MFN and MFN tariffs experienced more rapid acceleration in Chinese imports after 2000.

²² For comparability with subsequent columns, we report these models as reduced form OLS rather than 2SLS regressions.

²³ By contrast, the expected impact of rising predicted Chinese import competition using our primary strategy does not differ across periods. When an interaction term with the 1999–2007 period is added to the regression in column 3 of Table 3, its coefficient is very small and imprecisely estimated (coefficient 0.04, standard error 1.75), suggesting that the causal effect is comparable across the two periods.

TABLE 3—RESPONSE OF PATENTING TO ALTERNATIVE MEASURES OF TRADE EXPOSURE, 1991–2007. DEPENDENT VARIABLE: RELATIVE CHANGE OF NUMBER OF CORPORATE PATENTS BY US-BASED INVENTORS: OLS AND OLS REDUCED FORM ESTIMATES

	US imports from China (OLS)		Third country imports from China (OLS reduced form)		Normalized trade relations tariff gap (OLS reduced form)		MFA quotas (OLS reduced form)	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Exposure variable	-1.17 (0.16)		-1.81 (0.69)		1.95 (0.43)		-3.28 (2.22)	
Exposure variable × 1999–2007 period		-0.71 (0.33)		-1.63 (0.74)	-1.85 (0.59)	-0.92 (0.56)	-2.70 (1.39)	-2.27 (2.11)
Full controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Industry FEs		Yes		Yes		Yes		Yes
<i>Descriptives for exposure variable in 1999–2007 period</i>								
Mean		6.46		5.62		30.00		1.47
Standard deviation		(14.34)		(9.17)		(14.31)		(4.49)
Impact of 1σ exposure		-10.15		-14.91		-13.11		-10.21
<i>Share of manufacturing patents in four-digit industries with nonzero exposure, year 1999</i>								
		100%		100%		100%		4.9%

Notes: $N = 8,271$, except $N = 495$ in columns 7–8, which include only the three-digit sectors that comprise at least one MFA-affected four-digit industry. Every regression comprises two stacked first differences 1991–1999 and 1999–2007 and includes the full set of controls from model 3h in Table 2. Data on average fill rates of MFA quotas in 1999 and on normalized trade relations tariff gaps are based on Pierce and Schott (2016) and measured in percentage points. All models are weighted by a firm's US-inventor patents averaged over the start and end of a period. Standard errors are clustered on four-digit SIC industries.

(second row, column 5). Paralleling earlier estimates, column 6 introduces industry fixed effects into the model. The coefficient of interest in this demanding specification (tariff gap \times post) remains negative and is marginally statistically significant.

In columns 7 and 8, we employ the instrument of Bloom, Draca, and Van Reenen (2016), which corresponds to the share of industry imports subject to MFA quotas prior to the 2005 MFA termination.²⁴ As with the tariff gap instrument, we focus on the contrast between the pre-removal period of 1991–1999 and the removal period of 1999–2007. Following Bloom, Draca, and Van Reenen (2016), we analyze the impact of the removal of MFA quotas exclusively for the subsample of firms operating in MFA-affected industries. This reduces our sample from 8,271 to 495 observations, and the affected sectors account for just 4.9 percent of manufacturing patenting in 1999 (bottom row of Table 3). Nevertheless, the column 7 estimate finds a negative association between MFA quotas and US patent growth during 1991–1999, which becomes (marginally) significantly more negative in the 1999–2007 period when import quotas are phased out. When four-digit industry fixed effects are added in column 8, the point estimate remains comparable but is less precisely estimated.

The second-to-last row of Table 3 shows that for all three sets of instruments employed, a one standard deviation increase in the trade exposure instrument is estimated to reduce patenting by 10 to 15 log points. Thus, whether we exploit

²⁴ The MFA quota fill rate data come from Pierce and Schott (2016).

the observed growth in US imports from China, the supply-driven surge in China's exports to the United States and other high income countries, the reduction in China-US trade policy uncertainty following the grant of PNTR, or the removal of import quotas in MFA-affected industries, we find qualitatively and quantitatively similar deterrent effects on patenting by import-competing US firms.

As detailed in online Appendix Tables A5 and A6, our key finding that rising import competition reduces patenting by US firms is robust to variations of our basic approach, including restricting the analysis to only patents granted within six years of the application date, excluding patents from the computer or chemical technology classes, using alternative firm weights (patent citations, R&D expenditure, sales, no weights), and assigning Compustat firms to corresponding industries based on historical industry codes or using fractional industry assignments according to firms' sales shares across the industries in which they operate.²⁵

A limitation of our firm-level analysis is that the analytic sample excludes patents granted to firms that are *not* listed in Compustat. We overcome this limitation by estimating the effect of import competition on patenting at the *technology class* level. Exploiting the fact that every patent is assigned to a specific technology class, we calculate the import exposure facing each technology class by using the implicit mapping of patent classes to industries provided by our Compustat-patent matched data. Fitting (3) to patent counts at the technology class level, we find that technology classes with greater growth in import penetration during our window of 1991–2007 experienced a substantial relative decline in total *corporate* patents (see online Appendix Table A7). Yet, no such negative effect is detected for patenting by noncorporate entities (e.g., government and universities). Our findings are thus specific to private sector innovation and are unlikely to reflect a spurious correlation between declining innovation opportunities and rising trade exposure.

D. Which Firms Are Most Affected?

Our findings contrast with Bloom, Draca, and Van Reenen (2016), who report a positive effect of exposure to the removal of MFA quotas on patenting by European firms. These opposite-signed effects for the United States and Europe are not intrinsically at odds: as Aghion et al. (2005) emphasize, a marginal increase in competition may either spur or hinder innovation depending on the initial competitiveness of a market.²⁶ Our findings could imply that US manufacturing industries largely locate on the downward-sloping leg of the Aghion et al. (2005) inverted U-shaped locus, where greater competition diminishes innovation.

To explore heterogeneous effects of competition on firm patenting, Table 4 splits US firms within detailed industries into groups according to four metrics: sales per worker, capital per work, profit over capital (ROI), and debt to equity. The odd-numbered columns estimate the effect of import competition on patenting for firms that outperform their respective industry's mean in terms of

²⁵ We also find that import competition has little impact on firm industry representation: conditional on survival, Compustat firms do not appear to change industry in response to rising China competition.

²⁶ It is further possible that the shock from Chinese trade competition may be less pronounced in Europe due to a more balanced trade relationship with China. Dauth, Findeisen, and Suedekum (2014) estimate that the labor market effects of Chinese trade for Germany, for instance, have been much less severe.

TABLE 4—EFFECT OF CHINESE IMPORT COMPETITION ON PATENTING 1991–2007: SAMPLE SPLITS BY INITIAL FIRM SALES/WORKER, CAPITAL/WORKER, ROI, AND DEBT/EQUITY. DEPENDENT VARIABLE: RELATIVE CHANGE IN PATENTS

	Firm labor productivity and capital intensity				Firm profitability and leverage			
	Sales/worker		Capital/worker		Profit/capital (ROI)		Debt/equity	
	> Avg (1)	≤ Avg (2)	> Avg (3)	≤ Avg (4)	> Avg (5)	≤ Avg (6)	≤ Avg (7)	> Avg (8)
ΔUS industry exposure to Chinese imports	−1.11 (0.79)	−2.32 (1.38)	−0.01 (1.01)	−2.41 (0.55)	−0.86 (0.57)	−2.06 (0.72)	−0.86 (0.40)	−2.92 (1.47)
Test for equal coeff.	$p = 0.522$		$p = 0.034$		$p = 0.192$		$p = 0.149$	
Mean outcome variable	27.27	3.29	29.80	7.83	20.03	7.58	19.18	14.23
Observations	1,348	2,738	1,529	2,560	1,887	2,493	3,154	766

Notes: Every regression comprises two stacked first differences, 1991–1999 and 1999–2007, and includes the full set of controls from model 3h in Table 2. Columns 1–2, 3–4, 5–6, and 7–8 split the firm sample into firms whose sales per employee, capital per employee, ROI, or debt-to-equity ratio is above/below the patent-weighted industry average in the start-of-period year. All models are weighted by a firm’s US-inventor patents averaged over the start and end of a period. Standard errors are clustered on four-digit SIC industries.

higher productivity, capital intensity, profitability, and lower indebtedness. The even-numbered columns present analogous regressions for the complementary samples. All four sample splits qualitatively confirm that firms with an initially weaker competitive position are more likely to be hurt by foreign competition, and this heterogeneity in treatment effects is statistically significant for the sample split by capital intensity. In Table A9 of the online Appendix, we extend the results in Table 4 by redefining the dependent variable to be the change in patents relative to the change in firm size (measured as sales or employment, either for the United States or globally). We find that for the initially weaker firms—when measured by capital intensity or profitability—the trade-induced contraction in patenting is *not* simply a by-product of a reduction in firm scale. Instead, these firms experience a significant decline in patenting relative to sales and employment. Our analysis of firm heterogeneity suggests that rising import competition could, by culling the ranks of weaker incumbents, catalyze the reallocation of sales and profits to the stronger firms that remain. While it does not definitively explain why results differ between the United States and Europe, it does support the notion that the effect of increased competition on innovation depends on initial conditions.²⁷

IV. Discussion

We provide a comprehensive analysis of how the surge in import competition from China affects both the inputs and outputs of US innovation activities. Accounting for the important confounding effects of industry pre-trends, we demonstrate that

²⁷ In complementary analyses, Gutiérrez and Philippon (2017) find that the sensitivity of capital investment to changes in industry import competition is greater in smaller relative to larger US manufacturing enterprises, and Aghion et al. (2018) find that patenting by French manufacturing firms responds more positively to *export* demand shocks in more productive relative to less productive establishments.

the negative impact of Chinese import competition on innovation at the firm and technology class level is robust to a wide range of regression specifications and identification strategies, that it is evident on both output (patenting) and input (R&D expenditure) margins, and that it applies to private sector patent production but not patenting by universities or government, consistent with a competitive market response. These findings do not constitute a welfare analysis of the impact of competition on domestic innovation since they do not capture, for example, dynamic responses that may ultimately include the culling of weaker firms and the entry of robust new innovators. Nevertheless, they demonstrate that the innovation response of US firms more exposed to rising market competition from China has been substantially and unambiguously negative.

REFERENCES

- Acemoglu, Daron, David Autor, David Dorn, Gordon H. Hanson, and Brendan Price.** 2016. "Import Competition and the Great US Employment Sag of the 2000s." *Journal of Labor Economics* 34 (S1 Pt. 2): S141–98.
- Aghion, Philippe, Antonin Bergeaud, Matthieu Lequien, and Marc J. Melitz.** 2018. "The Impact of Exports on Innovation: Theory and Evidence." National Bureau of Economic Research Working Paper 24600.
- Aghion, Philippe, Nick Bloom, Richard Blundell, Rachel Griffith, and Peter Howitt.** 2005. "Competition and Innovation: An Inverted-U Relationship." *Quarterly Journal of Economics* 120 (2): 701–28.
- Arora, Ashish, Sharon Belenzon, and Andrea Pataconi.** 2018. "The Decline of Science in Corporate R&D." *Strategic Management Journal* 39 (1): 3–32.
- Arrow, Kenneth J.** 1962. "Economic Welfare and the Allocation of Resources for Invention." In *The Rate and Direction of Inventive Activity: Economic and Social Factors*, edited by Richard Nelson, 609–11. Princeton: Princeton University Press.
- Autor, David H., David Dorn, and Gordon H. Hanson.** 2013. "The China Syndrome: Local Labor Market Effects of Import Competition in the United States." *American Economic Review* 103 (6): 2121–68.
- Autor, David, David Dorn, Gordon H. Hanson, Gary Pisano, and Pian Shu.** 2017. "Foreign Competition and Domestic Innovation: Evidence from U.S. Patents." National Bureau of Economic Research Working Paper 22879.
- Autor, David, David Dorn, Gordon H. Hanson, Gary Pisano, and Pian Shu.** 2020. "Foreign Competition and Domestic Innovation: Evidence from US Patents: Dataset." *American Economic Review: Insights*. <https://doi.org/10.1257/aeri.20180481>.
- Autor, David H., David Dorn, Gordon H. Hanson, and Jae Song.** 2014. "Trade Adjustment: Worker-Level Evidence." *Quarterly Journal of Economics* 129 (4): 1799–1860.
- Balsmeier, Benjamin, Mohamad Assaf, Tyler Chesebro, Gabe Fierro, Kevin Johnson, Scott Johnson, et al.** 2018. "Machine Learning and Natural Language Processing on the Patent Corpus: Data, Tools, and New Measures." *Journal of Economics and Management Strategy* 27 (3): 535–53.
- Bartelsman, Eric, John Haltiwanger, and Stefano Scarpetta.** 2013. "Cross-Country Differences in Productivity: The Role of Allocation and Selection." *American Economic Review* 103 (1): 305–34.
- Bernard, Andrew B., J. Bradford Jensen, and Peter K. Schott.** 2006. "Survival of the Best Fit: Exposure to Low-Wage Countries and the (Uneven) Growth of U.S. Manufacturing Plants." *Journal of International Economics* 68 (1): 219–37.
- Bessen, James, and Robert M. Hunt.** 2007. "An Empirical Look at Software Patents." *Journal of Economics and Management Strategy* 16 (1): 157–89.
- Bloom, Nicholas, Mirko Draca, and John Van Reenen.** 2016. "Trade Induced Technical Change? The Impact of Chinese Imports on Innovation, IT and Productivity." *Review of Economic Studies* 83 (1): 87–117.
- Bloom, Nicholas, Paul M. Romer, Stephen J. Terry, and John Van Reenen.** 2014. "Trapped Factors and China's Impact on Global Growth." National Bureau of Economic Research Working Paper 19951.
- Boldrin, Michele, and David K. Levine.** 2013. "The Case Against Patents." *Journal of Economic Perspectives* 27 (1): 3–22.

- Chen, Cheng, and Claudia Steinwender.** 2019. "Import Competition, Heterogeneous Preferences of Managers, and Productivity." National Bureau of Economic Research Working Paper 25539.
- Cohen, Wesley M.** 2010. "Fifty Years of Empirical Studies of Innovative Activity and Performance." In *Handbook of the Economics of Innovation*, Vol. 1, edited by Bronwyn H. Hall and Nathan Rosenberg, 129–213. Amsterdam: Elsevier.
- Dasgupta, Partha, and Joseph Stiglitz.** 1980. "Industrial Structure and the Nature of Innovative Activity." *Economic Journal* 90 (358): 266–93.
- Dauth, Wolfgang, Sebastian Findeisen, and Jens Suedekum.** 2014. "The Rise of the East and the Far East: German Labor Markets and Trade Integration." *Journal of the European Economic Association* 12 (6): 1643–75.
- Feenstra, Robert C., and Gordon H. Hanson.** 1999. "The Impact of Outsourcing and High-Technology Capital on Wages: Estimates for the U.S., 1979–1990." *Quarterly Journal of Economics* 114 (3): 907–40.
- Gilbert, Richard.** 2006. "Looking for Mr. Schumpeter: Where Are We in the Competition–Innovation Debate?" *Innovation Policy and the Economy* 6: 159–215.
- Gordon, Robert J.** 2012. "Is US Economic Growth Over? Faltering Innovation Confronts the Six Headwinds." National Bureau of Economic Research Working Paper 18315.
- Gutiérrez, Germán, and Thomas Philippon.** 2017. "Declining Competition and Investment in the U.S." National Bureau of Economic Research Working Paper 23583.
- Hall, Bronwyn H., Adam B. Jaffe, and Manuel Trajtenberg.** 2001. "The NBER Patent Citation Data File: Lessons, Insights and Methodological Tools." National Bureau of Economic Research Working Paper 8498.
- Hall, Bronwyn H., Adam Jaffe, and Manuel Trajtenberg.** 2005. "Market Value and Patent Citations." *RAND Journal of Economics* 36 (1): 16–38.
- Handley, Kyle, and Nuno Limão.** 2017. "Policy Uncertainty, Trade, and Welfare: Theory and Evidence for China and the United States." *American Economic Review* 107 (9): 2731–83.
- Hart, Oliver D.** 1983. "The Market Mechanism as an Incentive Scheme." *Bell Journal of Economics* 14 (2): 366–82.
- Hashmi, Aamir Rafique.** 2013. "Competition and Innovation: The Inverted-U Relationship Revisited." *Review of Economics and Statistics* 95 (5): 1653–68.
- Helper, Susan, Timothy Krueger, and Howard Wial.** 2012. *Why Does Manufacturing Matter? Which Manufacturing Matters? A Policy Framework*. Washington, DC: Brookings Institution.
- Hombert, Johan, and Adrien Matray.** 2018. "Can Innovation Help U.S. Manufacturing Firms Escape Import Competition from China?" *Journal of Finance* 73 (5): 2003–39.
- Jaffe, Adam B., and Manuel Trajtenberg.** 2002. *Patents, Citations, and Innovations: A Window on the Knowledge Economy*. Cambridge, MA: MIT Press.
- Jorgenson, Dale W.** 2001. "Information Technology and the U.S. Economy." *American Economic Review* 91 (1): 1–32.
- Jorgenson, Dale W., Mun S. Ho, and Kevin J. Stiroh.** 2008. "A Retrospective Look at the U.S. Productivity Growth Resurgence." *Journal of Economic Perspectives* 22 (1): 3–24.
- Kogan, Leonid, Dimitris Papanikolaou, Amit Seru, and Noah Stoffman.** 2017. "Technological Innovation, Resource Allocation, and Growth." *Quarterly Journal of Economics* 132 (2): 665–712.
- Kueng, Lorenz, Nicholas Li, and Mu-Jeung Yang.** 2016. "The Impact of Emerging Market Competition on Innovation and Business Strategy." National Bureau of Economic Research Working Paper 22840.
- Li, Guan-Cheng, Ronald Lai, Alexander D'Amour, David M. Doolin, Ye Sun, Vette I. Torvik, Amy Z. Yu, and Lee Fleming.** 2014. "Disambiguation and Co-authorship Networks of the U.S. Patent Inventor Database (1975–2010)." *Research Policy* 43 (6): 941–55.
- Morrison, Greg, Massimo Riccaboni, and Fabio Pammolli.** 2017. "Disambiguation of Patent Inventors and Assignees Using High-Resolution Geolocation Data." *Scientific Data* 4: 170064.
- Moser, Petra.** 2016. "Patents and Innovation in Economic History." *Annual Review of Economics* 8: 241–58.
- Pierce, Justin R., and Peter K. Schott.** 2012. "A Concordance between Ten-Digit U.S. Harmonized System Codes and SIC/NAICS Product Classes and Industries." *Journal of Economic and Social Measurement* 37 (1–2): 61–96.
- Pierce, Justin R., and Peter K. Schott.** 2016. "The Surprisingly Swift Decline of US Manufacturing Employment." *American Economic Review* 106 (7): 1632–62.
- Shu, Pian, and Claudia Steinwender.** 2019. "The Impact of Trade Liberalization on Firm Productivity and Innovation." In *Innovation Policy and the Economy*, Vol. 19, edited by Josh Lerner and Scott Stern, 39–68. Chicago: University of Chicago Press.

- Ventura, Samuel L., Rebecca Nugent, and Erica R. H. Fuchs.** 2015. "Seeing the Non-stars: (Some) Sources of Bias in Past Disambiguation Approaches and a New Public Tool Leveraging Labeled Records." *Research Policy* 44 (9): 1672–1701.
- Xu, Rui, and Kaiji Gong.** 2017. "Does Import Competition Induce R&D Reallocation? Evidence from the U.S." International Monetary Fund Working Paper 17/253.

This article has been cited by:

1. Jesse Perla, Christopher Tonetti, Michael E. Waugh. 2021. Equilibrium Technology Diffusion, Trade, and Growth. *American Economic Review* **111**:1, 73-128. [[Abstract](#)] [[View PDF article](#)] [[PDF with links](#)]