Bombs, Brains, and Science

The Role of Human and Physical Capital for the Creation of Scientific

Knowledge

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Abstract

This paper analyzes the effects of human capital (HC) and physical capital (PC) for the productivity of science departments. To address the endogeneity of input choices I use two extensive but temporary shocks to the HC and PC of science departments. As HC shock I use the dismissal of mostly Jewish scientists in Nazi Germany. As PC shock I use the destruction of facilities by Allied bombings during WWII. In the short run, a 10 percent to HC lowered departmental productivity by about 0.21sd. A 10 percent shock to PC lowered departmental productivity by about 0.05sd in the short run. While the HC shock persisted until the end of my sample period (1980), departments experiencing a PC shock recovered very quickly (by 1961). Additional results show that the dismissal 'star scientists' was particularly detrimental, and that a fall in the quality of hires was an important mechanism for the persistence of the HC shock.

1 Introduction

Which inputs lead to successful scientific research? As for the production of goods, basic inputs for producing scientific knowledge are human and physical capital, i.e. scientists and their laboratories (Machlup, 1961). Understanding the relative role of these inputs is important for researchers, university administrators, and policy makers alike. At the moment, many countries such Brazil, South Korea, and especially China, are investing heavily in their university system (Kugler, 2011, Rhee, 2011, Wang et al. 2011). Should they hire outstanding scholars or construct new laboratories to achieve the highest return on their investment? The answers to these questions are particularly relevant in a world where ideas and scientific knowledge are seen as key drivers of technological progress and economic growth (Romer, 1990).

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Despite the significance of these issues we know little about the effects of different inputs for the production of scientific knowledge. As highlighted for firms by a large literature in industrial economics, the estimation of production functions is complicated because inputs are often chosen on the basis of unobservable productivity shocks (e.g. Ackerberg, et al. 2007). Estimating 'knowledge production functions' is similarly challenging. 'Star scientists' may be attracted by more productive departments and at the same time enhance the department's productivity. Similarly, high quality departments attract more funding for physical capital which further increases productivity. Obtaining unbiased estimates for the effects of human and physical capital is further complicated because good scientists attract funding for additional physical capital.

To overcome these difficulties I use two extensive, but temporary, shocks. As human capital shock I use the dismissal of mostly Jewish scientists in Nazi Germany and Austria between 1933 and 1940. As physical capital shock I use the destruction of science facilities by Allied bombings during WWII. The two shocks affected science departments very differently and therefore create ample variation. The dismissal of scientists reduced the faculty by up to 60 percent in some departments while others remained unaffected. Similarly, Allied bombings completely destroyed some departments while others were not affected.

I investigate how the two shocks affected department productivity in the short run, and whether they persisted in the long run. To analyze these effects I construct a new data set of all scientists in German and Austrian physics, chemistry, and mathematics departments at seven points in time between 1926 and 1980. The micro data contain more than 10,000 scientist-year observations with detailed productivity measures in top journals for each scientist. I use this data to construct productivity measures for all science departments between 1926 and 1980. I add data on the two shocks from detailed historical records of dismissals from German and Austrian universities in Nazi Germany and from archival material on bombing destruction during WWII.

Results show that both human and physical capital shocks had a negative effect on scientific output in the short run. A 10 percent shock to human capital lowered departmental productivity by about 0.21 standard deviations. A 10 percent shock to physical capital lowered productivity by about 0.05 of a standard deviation in the short run. As the two shocks only had a temporary effect on inputs, I can investigate their long-run effects. The human capital shock persisted in the long run and continued to have a negative impact on scientific output almost 50 years after the dismissals (in 1980). The physical capital shock, however, did not persist. The productivity of departments that had been bombed during WWII had recovered by 1961. By 1970, bombed departments even had slightly higher productivity than other departments; this suggests that bombed departments benefitted from upgrading during post-war reconstruction (these results are only significant in some specifications).

I show that the results are not driven by other changes that affected the German and Austrian university system which may have been correlated with the two shocks. In particular, the results are robust to controlling for post-war occupations zones (U.S., U.K., French, or

Soviet zones) and dropping East German and Austrian universities from the sample. The results are also robust to controlling for the creation of federal states after WWII. Similarly, controlling for changes at the university level, such as changes in university age and increased competition from newly founded universities does not affect the results. I also show evidence that changes at the city level are not driving my findings; the results are robust to controlling for bombing destruction at the city level, changes in the overall fraction of Jews in a city, investment in armament related industries by the Nazi government, and the erection of the 'iron curtain'.

Recent work has highlighted the contribution of 'star scientists' to the advancement of scientific knowledge (Azoulay, Zivin, and Wang, 2010). Many of the dismissed scholars were among the leaders of their profession; the dismissed include 11 Nobel Laureates such as the physicists Albert Einstein, Max Born, and Erwin Schrödinger and the chemists Fritz Haber and Otto Meyerhof. I can therefore investigate whether losing high quality scientists had particularly large effects on departmental productivity. I find that the dismissal of high quality scientists was associated with a larger productivity decline. Losing 'star' scientists was particularly detrimental. The loss of a scientist in the top 5th percentile of the quality distribution lowered long-run productivity by between 0.7 and 1.6 standard deviations (compared to an effect of 0.2 standard deviations for losing any scientist).

I then evaluate some possible mechanisms for the persistence of the human capital shock. I show that a reduction in department size after the dismissals only explains some of the productivity decline. A key mechanism for the persistence of the human capital shock was a permanent fall in the quality of hires, in particular after the dismissal of high quality scientists. As previous research has shown, localized productivity spillovers are unlikely to drive persistence in this context (Waldinger, 2011).

To my knowledge, no previous paper has analyzed the role of human and physical capital for the creation of scientific knowledge using exogenous variation in inputs. Existing empirical evidence has shown that scientific productivity of university departments is correlated with department size and research expenditure (Johnes, Taylor, and Francis, 1993). At the country level, patenting is significantly related to R&D manpower and spending (Furman, Porter, and Stern, 2002).¹

My findings also relate to several papers investigating the persistence of large economic shocks. Physical capital shocks, such as extensive bombings, usually dissipate relatively quickly (Davis and Weinstein, 2002, Brackman, Garretsen, Schramm, 2004, Miguel and Roland, 2011).²

¹A number of papers investigate other drivers of university productivity. University governance significantly affects how changes in funding affect research performance (Aghion et al., 2010). An increase of university funding increases the number of published papers but not their quality (Payne and Siow, 2003, Whalley and Hicks, 2011). At the level of individual scientists, National Institutes of Health (NIH) funding only has a limited impact on the research of marginal grant recipients (Jacob and Lefgren, 2011). Howard Hughes Medical Institute grants, however, which tolerate early failure and reward long-run success, increase the probability of publishing high-impact papers (Azoulay, Graff Zivin, and Manso, 2011).

²Other physical capital shocks, such as the massive infrastructure investment in the Tennessee valley between the 1930s and 1950s only had limited long-run effects that will likely disappear with time (Kline and Moretti, 2011).

Most human capital shocks, however, seem to persist in the long run. The extinction of the Jewish population in the Soviet Union by the German Army during WWII still affects city growth, per capita income, wages, and political outcomes today (Acemoglu, Hassan, Robinson, 2011) and reduces entrepreneurship and support for markets and democracy (Grosfeld, Rodnyansky, and Zhuravskaya, 2011). In Germany, the decline of the Jewish population during the Nazi era had persistent negative effects on education levels (Akbulut-Yuksel and Yuksel, 2011). In the present paper I analyze the persistence of human and physical capital shocks within the same framework for the first time. My results corroborate the findings of earlier papers that have separately analyzed human and physical capital shocks.

This paper also improves our understanding of Germany's decline as scientific superpower after WWII. At the beginning of the 20th century, German scientists were at the pinnacle of their profession (American Association for the Advancement of Science, 1941) and the leading German universities, especially Göttingen and Berlin, attracted large numbers of foreign scholars. Physicists like Arthur Compton (Nobel Prize, 1927) and Robert Oppenheimer from the United States, Leo Szilard and Eugene Wigner (Nobel Prize, 1961) from Hungary, Enrico Fermi (Nobel Prize, 1938) from Italy, and many Germans such as Werner Heisenberg (Nobel Prize, 1932), Max Born (Nobel Prize, 1954), and James Frank (Nobel Prize, 1925) had permanent or visiting positions in Göttingen during the 1920s (Dardo, 2004, p. 171).³ After WWII, German science declined massively and the United States became the dominant scientific force. This development is reflected in data on Nobel Prizes as shown in appendix Figure A1. Germany's decline may have been caused by a number of factors. The dismissal of some of the most prominent scientists (among them 11 Nobel laureates) and bombing destruction during WWII are obvious factors that I consider in this paper.

My estimates indicate that the dismissal of scientists reduced total output in affected German and Austrian science departments by 9,576 top journal publications between 1933 and 1980; a reduction of about 33.5 percent. In terms of citations weighted publications, affected German science departments lost 191,920 (34.6 percent) citations due to the dismissal. Dismissed scientists themselves produced 1,181 top journal publications receiving 32,369 citations. This shows that the effect of the dismissals on German science was much larger than the direct loss of publications and citations of dismissed scientists because the productivity loss in affected departments persisted at least until 1980.

WWII bombings of German science departments reduced total output of affected departments by 1,028 top journal publications between 1944 and 1980; a fall of about 5.7 percent. Citation weighted publications declined by 22,194 (6.4 percent).⁴ These calculations indicate that the dismissal of scientists in Nazi Germany contributed about 9 times more to the decline of German science than physical destruction during WWII.⁵

³Many of these illustrious scientists were later dismissed by the Nazi regime. Max Born and James Franck were dismissed from their professorships in Göttingen in 1933. Leo Szilard and Eugene Wigner who had moved to the University of Berlin and the Technical University of Berlin were also dismissed in 1933.

⁴The time periods for the calculation of percentage losses for dismissals and bombings are different because dismissals affected productivity from 1933 onwards, but bombings only affected productivity after 1943.

⁵For subject level results and details on these calculations see appendix section 8.3.

2 Human and Physical Capital Shocks

The production of scientific knowledge uses human and physical capital as main inputs (Machlup, 1961):

$$Y = f(H,K)$$

where Y measures output such as publications or citations in top journals, H measures human capital and K measures physical capital. University governance determines how human and physical capital are combined to produce scientific knowledge. Recent research has shown that more autonomous universities and those operating in a more competitive environment are better at converting funding increases into research output (Aghion et al., 2009).

Estimating knowledge production functions is challenging because inputs are often chosen on the basis of unobservable productivity shocks (Ackerberg et. al., 2007). 'Star scientists' may be attracted by highly productive departments and at the same time enhance the department's productivity. Similarly, investment in physical capital usually occurs in more productive universities. Establishing causality is therefore challenging. An additional problem is the separation of human and physical capital effects because good scientists like to work in universities with well equipped laboratories and attract funding for additional physical capital.

Even without these endogeneity concerns it would be challenging to directly estimate the production function of universities because of the difficulty to obtain information on physical capital of science departments over reasonably long time periods. I therefore use an indirect way to identify the importance of human and physical capital by investigating the persistence of large, and I argue exogenous, shocks to the human and physical capital of German and Austrian science departments. I estimate the short-run impact and the long-run persistence of these shocks as follows:

$$Output_{dt} = \beta_1 + \sum_t \beta_{2t} HCShock(1933-40)_d * Year_t + \sum_t \beta_{3t} PCShock(1942-45)_d * Year_t$$

$$+ \beta_4 DepartmentFE_d + \beta_5 YearFE_t + \varepsilon_{dt}$$
(1)

Where Output is a measure of a department d's research output in year t, such as total publications or citation weighted publications. HCShock(1933-40) measures the temporary shock to human capital during the dismissal of mainly Jewish scientists between 1933 and 1940. PCShock(1942-45) measures the temporary shock to physical capital during Allied bombings that occurred between 1942 and 1945. I describe both shocks in more detail below. The interactions of the shocks with year dummies (one for each of the 7 years between 1926 and 1980 for which I observe department productivity)⁷ allow me to investigate the persistence of the two shocks.⁸

⁶In fact, I have unsuccessfully tried to obtain consistent measures of physical capital for German and Austrian science departments for this paper.

⁷One of the 7 interactions will be the excluded category for each shock.

⁸Some science professors may have been killed by Allied bombings and thus bombings may have had a direct impact on human capital. In my data there is no evidence that the number of scientists who disappear from the sample between 1940 and 1950 is correlated with bombing destruction at the department level. In a regression

2.1 Human Capital Shock: The Dismissal of Scientists in Nazi Germany

As a human capital shock I use the dismissal of Jewish and 'politically unreliable' scientists by the Nazi government. Just over two months after the National Socialist Party seized power in 1933, the Nazi government passed the "Law for the Restoration of the Professional Civil Service" on the 7th of April of 1933. At once, all Jewish and 'politically unreliable' persons were dismissed from civil service positions in Germany. Anybody with at least one Jewish grandparent was to be dismissed. All civil servants therefore had to document their ancestry and those with Jewish grandparents were dismissed from service. Scientists of Jewish origin who had been civil servants since 1914 or who had lost a close family member in WWI were initially exempted from the dismissals. In 1935, however, the Reich citizenship laws (Reichsbürgergesetz) revoked the exemption and remaining scientists of Jewish origin were ultimately dismissed. The 1933 law was also used to dismiss civil servants with opposing political views. All members of the Communist Party, for example, were classified as 'politically unreliable' persons and were dismissed. The law was immediately implemented and resulted in a wave of dismissals and early retirements from the German universities.⁹

Overall, more than 1,000 academics were dismissed from the German universities. Shortly after the annexation of Austria on the 12th of March, 1938 the law was also implemented in Austrian universities. My calculations indicate that 15.0 percent of physicists, 14.1 percent of chemists, and 18.7 percent of mathematicians were dismissed between 1933 and 1940 (Table 1). Most dismissals from German universities occurred in 1933, immediately after the law was implemented. The small spike in dismissals in 1938 is driven by dismissals from Austrian universities.¹⁰

Many dismissed scientists were outstanding members of their profession. On average, they published more papers in top science journals and received more citations than scientists who were not dismissed. While 15.0 percent of physicists were dismissed, they published 23.8 percent of papers in top journals before 1933 and received 64 percent of the citations to papers published before 1933. In chemistry, 14.1 percent were dismissed but they published 22 percent of the papers in top journals, and received 23.4 percent of the citations. In mathematics, 18.7 percent were dismissed, but they contributed 31.0 percent of the top journal publications, and received 61.3 percent of the citations (Table 1). Moreover, 11 current or future Nobel laureates were

of the number of scientists who disappear from the sample in each department (for non-retirement reasons) on bombing destruction, the coefficient on destruction interacted with the 1950 dummy is -0.006 with a p-value of 0.43. The lack of a correlation of the bombings with the unexplained disappearance from the sample is not surprising because the total number of bombing casualties was relatively low. Historians are still debating about the exact number of bombing casualties but estimates range from 305,000 (United States Strategic Bombing Survey, 1945, p. 95) which is probably an underestimate to between 750,000 and one million (Frankland, 2005, p. 833) which may be too high. The German population at the beginning of WWII was about 69.3 million. Therefore, between 0.4 and 1.4 percent of the German population was killed by Allied bombings.

⁹For more details on the dismissal of professors and the consequences for Ph.D. students and contemporaneous effects on faculty peers see Waldinger (2010, 2012).

¹⁰Dismissals that occurred in German universities after 1933 affected researchers who had been exempted under the clause for war veterans or for who had taken up their position before 1914. Furthermore, some political dismissals occurred after 1933.

dismissed.

My data do not allow me to identify whether researchers were dismissed because of their Jewish origin or because of their political orientation. Historical studies, however, have shown that about 87 percent of the dismissed in chemistry (Deichmann, 2001), and 79 percent of the dismissed in mathematics (Siegmund-Schultze, 1998) were either Jewish or of Jewish decent. The remaining dismissals occurred for political reasons.

Most of the dismissed scientists emigrated and the majority of them obtained positions in foreign universities (Moser, Voena, and Waldinger, 2012). The main destinations were the United States, the United Kingdom, Turkey, the British Mandate of Palestine (later Israel), and Switzerland. A very small minority of the dismissed did not leave Germany and most of them died in concentration camps or committed suicide. Extremely few managed to stay in Germany and survive the Nazi regime, they were no longer allowed to use university laboratories and other resources. As a result, they did not contribute to the scientific output of their former departments.

The aggregate numbers hide the fact that German and Austrian universities were affected very differently by the dismissals. Even within a university there was a lot of variation across different departments. Some departments lost more than 60 percent of their personnel while others remained completely unaffected (Table 2). An example for the large variation in dismissals is the University of Heidelberg. It lost 60 percent of the mathematicians and almost 10 percent of chemists during the dismissals. In physics, however, not a single scientist was dismissed.

2.2 Physical Capital Shock: Allied Bombings of Universities in World War II

As physical capital shock I use destruction by allied bombings during WWII. At the beginning of the war in 1939, the Royal Air Force (RAF) concentrated bombings on military targets, such as the German fleet. Other targets were only bombed after the German invasion of the Low Countries and the bombing of Rotterdam by the Luftwaffe in May, 1940. To avoid the German antiaircraft defence the majority of bombing raids were flown under the cover of darkness; this made targeting extremely difficult. During this period, the RAF mainly targeted oil reservoirs, railway lines in the Ruhr area, aircraft factories, aerodromes, U-boat shipyards, and ports.

In the autumn of 1940, the RAF flew the first "area attacks" on German cities to "affect the morale of the German people" (Webster and Frankland, 1961 p. 156) and to "concentrate the maximum amount of damage in the centre of town" (Peirse, 1940). The first "area attack" was flown by 134 RAF bombers on December 16th, 1940, and targeted the inner city of Mannheim in the south of Germany. The attack was a response to the devastation of Coventry by the German Luftwaffe during the autumn of 1940.

During 1941, the RAF increased the number of small scale "area attacks" on German cities. Most of these attacks, however, did not cause any large destruction as only about 20 percent of bombers managed to navigate within five miles of their destination, even less managed to hit the target. As a result, the smallest potential targets were whole towns (Frankland, 2005, Webster and Frankland, 1961, vol. 1, p.156, p.257). Even these were often missed. A bombing raid of Karlsruhe and Stuttgart on October 1st, 1941, for example, not only hit the two targets but also 25 other cities such as Aachen, Chemnitz, or Würzburg, several hundred kilometers away (Webster and Frankland, 1961, vol. 1, p. 185).

The appointment of Sir Arthur Harris as head of "Bomber Command" on February 23rd, 1942, and the "Area Bombing Directive" that was issued a week earlier caused an intensification of the bombing campaign as more planes were deployed in each raid. An example is the first 1,000 bomber attack against Cologne on May 30th, 1942. The raid damaged about a third of Cologne's surface area (Hohn, 1991 p.12, Webster and Frankland, 1961, vol. 1, p. 340, pp. 402-410). To maximize destruction of inner cities the RAF used incendiary bombs that caused great fires in bombed cities. The introduction of heavy bombers (in particular the Lancaster bomber that was gradually introduced after March, 1942), the use of radar and radar-like devices (introduced in March, 1942), and the deployment of Pathfinder target marking planes (first used in January, 1943) increased the precision and efficiency of bombings.

In January, 1943, the United States Army Air Force (USAAF) entered the bombing campaign against Germany.¹² While the British continued to fly night time raids and in particular "area attacks" against inner cities, the USAAF mostly attacked during the day and bombed strategically important targets such as the German aircraft and ball bearings industries.

The bombing of targets in Germany intensified in 1944 with the introduction of the "double blow" tactic. Two, or later three, bombing attacks over short time periods increased the efficacy of incendiary bombs. The increased air supremacy of the Allied forces further facilitated the bombings. Towards the end of the war the bombardments were extended to smaller cities that had been spared in previous attacks.

Overall, about 1.35 million tons of bombs were dropped over German territory during WWII. Data on monthly bomb loads show an almost continuous increase between 1940 and 1945, with particularly large increases in the last years of the war (Figure 1).

¹¹The previous 107 attacks on Cologne had involved at most 40 planes each.

¹²While the first months of the US air offensive against Germany were only moderately successful, the effectiveness of USAAF operations greatly improved towards the end of 1943. These changes were achieved by an increase in the production of bombers, by the introduction of long-range Mustang fighters which gave bomber formations added protection, and by the establishment of U.S. air bases in Italy.

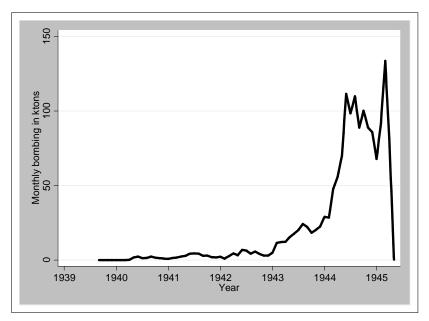


Figure 1: Allied bombings

Note: The figure shows the monthly amount of bombs (in tons), that was dropped on German territory during WWII. Data source: Webster and Frankland (1961), Annex.

Allied bombings, in particular during the last two years of WWII, completely destroyed about 18.5 percent of homes in what later became the Federal Republic of Germany (Hohn, 1991, p. 59). As area bombings targeted the inner cities of all larger cities they were most heavily affected.

Universities were never listed as targets in any of the Allied bombing directives and similar documents. Nonetheless, many university facilities were destroyed because bombs could not be precisely aimed until the end of WWII. Because of the targeting problems bombs fell relatively randomly within cities and there was large variation in destruction across different university departments (Table 2). Targeting buildings of particular departments would have been impossible. Because many bombing raids involved the use of incendiary bombs, fires in affected buildings destroyed most of the scientific equipment and important manuscripts that had not been relocated to safer locations.¹³ The fires also destroyed many of the valuable private libraries that professors had assembled during their career.

As bombings intensified in 1944 most science departments were hit in 1944 and 1945. For some universities I obtain exact dates for the first and last bombing raid that destroyed university buildings. According to these data, the first raid occurred in June, 1941, and destroyed buildings at the Technical University of Aachen. Bombing raids that hit other universities continued and intensified until the end of WWII.

After the end of the war on May 8th, 1945 reconstruction of university buildings was initially hampered by a lack of funds and skilled craftsmen, missing supplies, and the devastation of

¹³Many departments relocated material to safer town but even those were often bombed later on. The theoretical physics section of the University of Bonn, for example, moved equipment to Hildburghausen in Thuringia but the temporary site was bombed towards the end of the war (van Rey, 1995 p. 36).

many German cities. Most universities enlisted students to help with reconstruction. The universities of Bonn, Karlsruhe, and Hannover, for example, required up to 1,000 reconstruction service hours from its students until 1949 (van Rey, 1995 p. 42, Hoepke, 2007 p. 137, Wolters, 1950 pp. 123-129). Most universities completed reconstruction by the end of the 1950s but some work lasted until the 1960s (Hoepke, 2007 p. 139, Technische Universität Dresden 1996, pp. 18-36).

3 Panel Data Set of German Science Departments

3.1 Scientists in German and Austrian Universities from 1926 to 1980

To evaluate the impact and persistence of the two shocks on the productivity of science departments I construct a new panel data set covering physicists, chemists, and mathematicians at German and Austrian universities. The data come from "Kürschners Deutscher Gelehrtenkalender" that has been published since the 1920s in 5 to 10 year intervals. Data from volumes published in 1926, 1931, 1940/41, 1950, 1961, 1970, and 1980 allow me to construct complete faculty rosters for science departments at these seven points in time, spanning 54 years. From each volume I extract all scientists who were chaired professors, extraordinary professors, or 'Privatdozenten' (the first position in the German university system with the right to give lectures). To obtain a sample that is comparable over time, I focus on scientists in all 35 German or Austrian universities that existed in 1926 and remained on German (both FRG and GDR) or Austrian territory after 1945 (see Table 2 for a listing of the universities in my sample). As I obtain data on physics, chemistry, and mathematics departments in each university, the final data set contains 105 science departments. The data appendix provides more details on the scientists data.

The micro data include 5,716 scientists (2,456 chemists, 2,000 physicists, and 1,260 mathematicians) with 10,387 person-year observations (4,605 in chemistry, 3,594 in physics, and 2,188 in mathematics). The number of scientists in German and Austrian universities increased massively after 1926. The exception is 1941 (and 1950 in chemistry), the first data point after the dismissal of scientists (Figure 2).

¹⁴Privatdozenten are comparable to junior faculty at U.S. universities. Extraordinary professors are comparable to associate professors and chaired professors are comparable to full professors in the U.S. system.

¹⁵This excludes three universities (the University of Breslau, now Wroclaw, the Technical University of Breslau and the University of Königsberg, now Kaliningrad) that were on German territory in 1926 but became part of Poland or the Soviet Union after 1945. As the "Gelehrtenkalender" no longer listed researchers from these three universities after 1945 they cannot be included in the long-term analysis. Most scientists from these universities relocated to universities that continued to be in Germany after 1945.

¹⁶Scientists in universities located in the GDR were not listed in the "Gelehrtenkalender" in 1980.

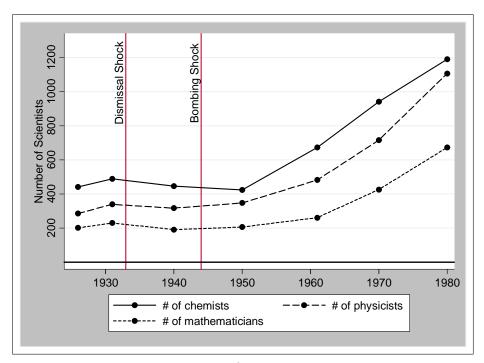


Figure 2: Number of scientists 1926-1980

Note: The figure reports the number of scientists in German and Austrian universities in chemistry, physics, and mathematics at each of the 7 points in time covered by the data. Data come from Kürschners Deutscher Gelehrtenkalender (see text for details).

3.2 Productivity Measures for German and Austrian Science Departments

To measure productivity at the department level I first obtain publications and citation weighted publications for each scientist in the sample. The publication data are downloaded from the "ISI Web of Science" and include all top journals for German and Austrian scientists. As journal rankings changed over time, I download both historical and current top journals.

The list of historical top journals includes all science journals published in Germany and are covered by the Web of Science between 1920 and 1944.¹⁷ I add a number of general science journals (e.g. Nature and Science) and historically relevant international field journals (e.g. Acta Mathematica) that were outlets for German scientists in the 1920s and 30s. I augment the list of historical top journals with a set of current top journals that I obtain from commonly used journal rankings. See Table A1 for the complete list of top journals used in the analysis and the data appendix for additional information on the journal data.

I download all articles published between 1920 and 1985 in any of the journals in my list. 18 I

¹⁷Historical journals with coverage in the Web of Science were the top journals at their time because Thomson Scientific digitized only the most cited journals for the period 1900 to 1944. See http://wokinfo.com/products_tools/backfiles/cos/ for further information on the digitization of historical top journals.

¹⁸A few top journals, such as Physical Review Letters, were founded after 1920. For these journals I download all articles after their creation. The changing pool of journals does not affect my findings as all regressions include year fixed effects.

can therefore calculate the number of top journal publications (and citation weighted publications) for each scientist. The Web of Science data only include the last name and the initial of the first name (or two initials if the author uses two first names) for each author. Most German scientists have distinct last names that are also different from most foreign names. In the rare cases that the last name - first initial combination does not uniquely identify a scientist in my data I split (citation weighted) publications according to the number of scientists with the same last name - first initial combination. Table A2 shows the most cited scientists in my data. Most of these scientists are very well known in the scientific community. This indicates that the productivity measures carry meaningful information. Interestingly, Johann von Neumann who later emigrated to the United States is the most cited mathematician in the data.

To measure department productivity for each of the seven points in time I add individual productivity measures within departments. Individual productivity is measured using a five year window around the relevant year. Albert Einstein's individual productivity measure for 1926, for example, is the sum of his publications between 1923 and 1927. I then sum these individual productivities within departments. Say a department had three scientists with individual productivities equal to 1, 2, and 3; total department productivity would then be 1+2+3=6. The Web of Science data not only include publications but also the number of times each article was subsequently cited in any journal covered by the Web of Science. I can therefore construct an additional productivity measure based on citation weighted publications. ²¹

Publication and citation patterns are very different across the three subjects. To ensure comparability across subjects I normalize total department productivity to have zero mean and unit variance in each subject. This also allows for easy interpretation of the estimated regression coefficients.

3.3 Data on Dismissals

I obtain data on dismissed scientists from a number of sources. The main source is the "List of Displaced German Scholars". It was compiled by the relief organization "Emergency Alliance of German Scholars Abroad" that had been founded by some dismissed scientists with the purpose of supporting other dismissed scholars to find positions in foreign universities. The list was published in 1937 and contained about 1,650 names of dismissed researchers. I extract all

¹⁹To reduce measurement error I focus on publications that correspond to the scientist's subject. For physicists, for example, I only use publications in top physics and general science journals that are outlets for physicists. I do not consider possible publications in chemistry journals. While this may underestimate the publications of some scientists it is preferable to counting large numbers of irrelevant publications because the last name - first initial combination does not perfectly identify scientists, especially across fields.

²⁰Publications are measured using an asymmetric window around the relevant year of the "Gelehrtenkalender"; i.e. productivity for the faculty in 1926 is measured with publications between 1923 and 1927 (instead of publications between 1924 and 1928). This asymmetry accounts for the delay in the publication of the "Gelehrtenkalender" as questionnaires for a certain volume had to be sent out and returned before publication. Using a symmetric window to compute average productivities does not affect the results.

²¹The citation weighted productivity measure is constructed as above by adding all citations to publications published in a five year window around the relevant year. For 1926, for example, I add all citations to articles published between 1923 and 1927. Citations are counted until today.

dismissed physicists, chemists, and mathematicians from the "List".

As the "List" was published before 1938 it did not include dismissals from Austrian universities. I consult the "Biographisches Handbuch der deutschsprachigen Emigration nach 1933 - Vol. II: The arts, sciences, and literature (1983)" to obtain dismissals from Austria. ²² This source also contains a few additional dismissals from German universities, for example, because dismissed scientists passed away before the "List of Displaced German Scholars" was compiled. The two sources together cover about 90 percent of all dismissals. I augment this information with data on a few additional dismissals from three secondary sources compiled by historians who have studied the dismissal of scientists in Nazi Germany. ²³

3.4 Data on Bombings of Science Departments

No existing data set covers bombing destruction of German and Austrian science departments. To measure department level bombing destruction, I therefore assemble a new data set from information in university archives. Bombed university institutes often provided detailed reports of destruction levels to obtain funds and materials for reconstruction. I construct departmental destruction using information from these reports that I obtain in a two-step process.

First, I contacted university archivists and asked them for information on destruction levels for all buildings used by physicists, chemists, and mathematicians. In some cases departments did not report destruction in percentages but instead gave verbal descriptions of bombing damages. I convert this information into percentage destruction using a rule outlined in the data appendix. For universities where archivists could not provide destruction information, I or my research assistant personally consulted the historical reports in the respective university archive to obtain the missing information.²⁴

To analyze the importance of measurement error I also construct a measure of bombing destruction at the university level. The data for this alternative measure come from information on university websites and a number of additional sources (Tietze, 1995, Phillips, 1983, Samuel and Thomas, 1949, Schneider, 1990, and Cheval, 1991). The data appendix provides additional details on the destruction data.

3.5 Data on Control Variables

To investigate the robustness of my findings to the inclusion of additional controls I obtain data on university age, the creation of nearby universities, the share of firms in armament related

²²The "Handbuch" was jointly compiled by the "Institut für Zeitgeschichte München" and the "Research Foundation for Jewish Immigration New York". I use information contained in Kröner (1983) to extract all dismissed scientists from the "Handbuch".

²³Dismissed chemists are contained in Deichmann (2001), dismissed physicists in Beyerchen (1977) and dismissed mathematicians in Siegmund-Schultze (1998).

²⁴This two-step method did not yield department level destruction data for the University of Darmstadt. I therefore use university level destruction for this university.

industries in 1933, the fraction of Jews at the city level in 1933, and the distance to the 'iron curtain'. Further information on the control variables can be found in the data appendix.

Combining the data from all sources I obtain a panel data set of German and Austrian science departments that I observe at seven points in time (1926, 1931, 1940, 1950, 1961, 1970, and 1980). The data include different measures of department productivity, information on dismissal and bombing shocks, and time-varying control variables.

4 Persistence of Human and Physical Capital Shocks

4.1 Main Results

I investigate the persistence of the human and physical capital shocks by estimating equation (1). First, I analyze how the dismissal shock affected the productivity of science departments. As dismissals occurred between 1933 and 1940, with most dismissals happening in 1933, the first post-dismissal observation is 1940. Regression results indicate that between 1931 (the last data point before the dismissals) and 1940 productivity in departments with dismissals fell significantly compared to departments without dismissals (Table 3, column 1). Coefficients on the interactions with subsequent years indicate that this effect persisted until 1980. Controlling for subject times year fixed effects to allow for differential productivity trends in the three subjects does not affect the results (column 2). Further controlling for occupation zone (U.S. zone, U.K. zone, French zone, Soviet zone) times post-1945 dummies has a negligible effect on estimated coefficients but lowers standard errors (column 3). The estimates imply that the dismissal of one scientist lowered department productivity, even the long run, by between 0.17 and 0.28 standard deviations (column 3).

I next analyze how the bombing shock affected productivity. As Allied bombings intensified towards the end of the war, the first post-bombing observation is 1950. The estimates indicate that productivity of departments that were bombed during WWII declined between 1941 and 1950 compared to other departments. This effect is only significant at the 5 percent level if I control for subject times year fixed effects and occupation zone times post-1945 dummies (Table 3, columns 4 to 6). The estimates imply that the destruction of 10 percent of department buildings lowered productivity by 0.05 standard deviations. Productivity recovered very quickly after WWII. Already in 1961, there was no significant difference between departments that were bombed during WWII and other departments. By 1970, departments that were bombed even performed slightly better than other departments. While these results are only significant at the 10 percent level in two of the three specifications they suggest that upgrading during reconstruction may have had a small positive effect on the productivity of bombed departments in the longer-run.

Jointly estimating long-run effects of the bombing and dismissal shocks leads to very similar conclusions (Table 3, columns 7 to 9). To investigate whether the bombing results are driven

by the destruction of department buildings or by more general destruction at the city level, I add interactions of year dummies with city level destruction to the regression. The results do not change substantially (Table 3, column 10). This indicates that productivity is primarily driven by department facilities and not by more general city level effects.

To investigate the relative magnitude of the two shocks I plot the effect of 10 percent shocks to the human and physical capital of a hypothetical science department. For this exercise I scale coefficients and standard errors as reported in column (10) of Table 3 to reflect a 10 percent shock to both human and physical capital.²⁵

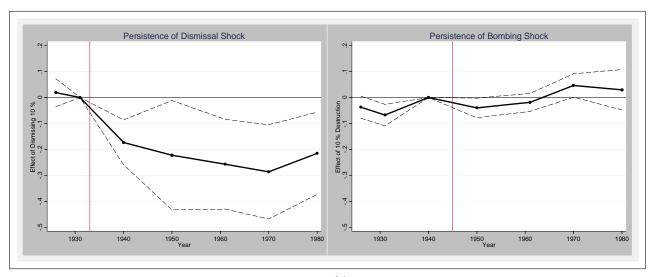


Figure 3: Persistence of 10% shocks - publications

Note: The figure plots scaled regression coefficients and 95 percent confidence intervals obtained from the estimation of equation (1) as reported in column (10) of Table 3. Point estimates and confidence intervals are scaled to reflect a 10 percent shock to both human and physical capital. The dependent variable is the total number of publications in department d and year t. The left hand panel reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies. The excluded year is 1931; the last data point before the dismissals. The right hand panel reports coefficients on the interaction of percentage destruction (by Allied bombings between 1941 and 1945) with year dummies. The excluded year is 1940; the last data point before the bombings.

The exercise reveals that a 10 percent shock to human capital had a larger effect on productivity than a 10 percent shock to physical capital, even in 1950 (Figure 3).²⁶ While the human capital shock persisted for more than 40 years, until 1980, the physical capital shock dissipated quickly and even had a small positive effect on departmental productivity in 1970. The figure also shows that pre-trends are unlikely to drive the human capital results. For the bombing shock the coefficient on the interaction of the 1931 dummy with WWII destruction is negative and significant. This indicates that the productivity of departments that were bombed

²⁵Average department size in 1931 was 10.15. A 10 percent shock to human capital therefore corresponds to losing 1.015 scientists. I therefore multiply the dismissal coefficients and standard errors reported in column (10) of Table 3 by 1.015. As departmental destruction is already measured in percentages, I multiply the bombing coefficients and standard errors reported in column (10) of Table 3 by 10.

 $^{^{26}}$ In 1950, the p-value of a test of the Null hypothesis (1.015*coefficient # dismissed*1950) = (10*coefficient % destruction*1950) has a p-value of 0.097. For all later years, the effect of a 10 percent human capital shock is significantly larger (at the 5 percent level) than a 10 percent physical capital shock (p-values between 0.001 and 0.012).

during WWII improved before the bombings occurred. This pre-trend suggests that the effect of bombing destruction may be underestimated in 1950. I show below, that controlling for other factors that may have affected productivity reduces the magnitude and significance of the pre-war bombing coefficients while leaving the post-war results unchanged.

In the previous results productivity was measured as the sum of publications in top journals (normalized to have a mean of zero and a standard deviation of one). Alternatively, productivity could be measured by citation weighted publications. This measure multiplies publications by the number of subsequent citations in articles that were published in any journal covered by the Web of Science (again normalized to have a mean of 0 and unit variance). The citations quantify the quality of each published paper using valuations of the entire scientific community.

The alternative productivity measure yields very similar results (Table 4). The dismissal of one scientist reduced productivity, even in the long run, by between 0.16 and 0.22 standard deviations (column 10). Bombings had a small negative, but insignificant, effect in 1950. By 1970, bombed departments even performed slightly better (significant at the 10 percent level). Figure 4 plots the long-run effects of 10 percent shocks to human and physical capital corresponding to the estimates reported in column (10) of Table 4. The dismissal shock had a larger effect than the bombing shock and was much more persistent.²⁷

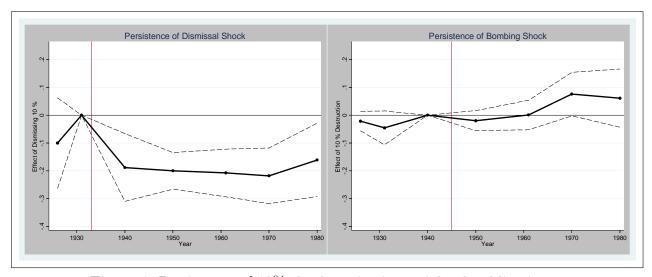


Figure 4: Persistence of 10% shocks - citation weighted publications

Note: The figure plots scaled regression coefficients and 95 percent confidence intervals obtained from the estimation of equation (1) as reported in column (10) of Table 4. Point estimates and confidence intervals are scaled to reflect a 10 percent shock to both human and physical capital. The dependent variable is the total number of citation weighted publications in department d and year t. The left hand panel reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies. The excluded year is 1931; the last data point before the dismissals. The right hand panel reports coefficients on the interaction of percentage destruction (by Allied bombings between 1941 and 1945) with year dummies. The excluded year is 1940; the last data point before the bombings.

 $^{^{27}}$ The test of the Null hypotheses (1.015*coefficient # dismissed*year dummy) = (10*coefficient % destruction*year dummy) has p-values between 0.000 and 0.011.

4.2 Robustness of Main Results

In the following, I show that the results are robust to the inclusion of additional control variables, to the adjustment of productivity for changes in the age-structure of a department, and to the use of different samples. I also investigate whether the way I define the two shocks variables or measurement error in the destruction variable are driving my findings.

Additional Controls

After WWII, federal states (Länder) became responsible for universities in West Germany. To investigate whether these changes were correlated with the shocks and therefore affect my results, I add interactions of federal state indicators with a post-1945 dummy to the regression.²⁸ Reassuringly, the results remain almost unchanged (Tables 5 and 6, column 1).

In a further robustness check, I add university age and its square to the regression. The specification allows me to rule out that the two shocks disproportionately affected older (or younger) universities that may have been on different productivity trends. The results are almost unaffected (Tables 5 and 6, column 2).

During the post-war period, in particular during the 1960s and 1970s, a number of new universities were founded in Germany and Austria. Increased competition from these universities may have influenced the productivity of established departments. To investigate whether my results are affected by these changes I include a time-varying control that measures the number of departments within 50 kilometers of each science department. The results remain unchanged (Tables 5 and 6, column 3).

After the Nazi government seized power in 1933, it invested heavily in rearmament (Tooze, 2006). As this investment was concentrated in a few industries it may have affected department productivity in cities with firms that benefitted from rearmament spending, either directly through spillovers from industry to universities or because the Allies may have targeted these cities during the bombing campaign. I therefore interact the share of firms in three armament related industries with year dummies and include these interactions as additional controls.²⁹ Reassuringly, the results are affected very little (Tables 5 and 6, column 4).

Departments with more Jewish scientists may have been located in cities with a higher fraction of Jewish residents. The disappearance of the Jewish population may have had long-lasting effects on these cities (e.g. Akbulut-Yuksel and Yuksel, 2011, Acemoglu, Hassan, Robinson, 2011, and Grosfeld, Rodnyansky, and Zhuravskaya, 2011) that could have affected the productivity of science departments in the long run. To investigate whether the results are affected by

²⁸The GDR was a centralized state and did not re-introduce federal states after WWII. I therefore include a joint indicator for all universities in East Germany (including the Technical University of Berlin that was located in West Berlin) in the regression. The federal states of Hamburg and Schleswig-Holstein had one university only and I therefore combine them with the adjacent state of Niedersachsen. The five Austrian universities are covered by a joint indicator. Including interactions of separate indicators for states with only one university with a post-1945 indicator hardly affects the results.

²⁹The three armament relevant industries are iron and steel production, mechanical engineering and vehicle construction, and chemicals. The shares of firms in these industries are measured in 1933 (1930 for Austrian cities). They are therefore determined before the Nazi government seized power and are therefore not endogenously affected by the two shocks.

these changes, I add the interaction of a post-1945 dummy with the fraction of Jews at the city level in 1933 to the regression.³⁰ If anything, the dismissal coefficients become slightly more negative (Tables 5 and 6, column 5). This indicates that the extinction of Jews from German and Austrian cities is not driving the results.

Finally, I investigate whether the two shocks were correlated with the geographic location of universities. Universities in cities that were closer to the 'iron curtain' may have suffered after 1945, because these cities experienced a decline in population growth after the division of Germany (Redding and Sturm, 2008). To investigate this concern I add the interaction of a post-1945 dummy with distance to the iron curtain to the regression. The results are unchanged (Tables 5 and 6, column 6).

Figure 5 shows the effect of 10 percent shocks including all controls. The figure looks very similar to the figure with the limited set of controls (Figure 3). This suggests that my findings are not driven by other factors that may have affected productivity. The figure also shows that the addition of controls reduces the magnitude and significance of pre-trends for the bombing results while leaving the effect in 1950 unchanged.

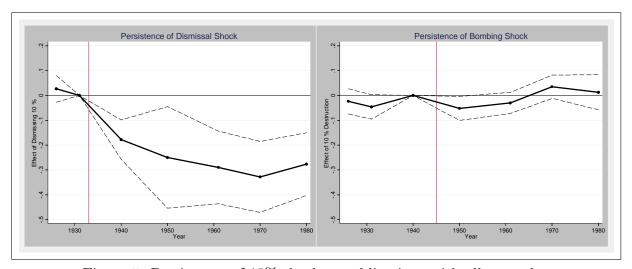


Figure 5: Persistence of 10% shocks - publications with all controls

Note: The figure plots scaled regression coefficients and 95 percent confidence intervals obtained from the estimation of equation (1) as reported in column (6) of Table 5. Point estimates and confidence intervals are scaled to reflect a 10 percent shock to both human and physical capital. The dependent variable is the total number of publications in department d and year t. The left hand panel reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies. The excluded year is 1931; the last data point before the dismissals. The right hand panel reports coefficients on the interaction of percentage destruction (by Allied bombings between 1941 and 1945) with year dummies. The excluded year is 1940; the last data point before the bombings.

Figure 6 shows the effect of 10 percent shocks for the citation-weighted productivity measure including all controls. It also looks similar to the figure with the limed set of controls (Figure 4). The small uptick in productivity in 1980 is less pronounced if I add additional controls to the regression.

³⁰For Austrian cities the fraction of Jews at the city level is measured in 1934 (Wien) and 1938 (Graz and Innsbruck) and thus before the annexation of Austria by Nazi Germany in 1938.

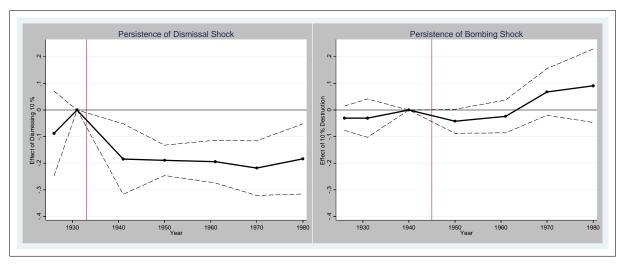


Figure 6: Persistence of 10% shocks - citation weighted publications with all controls

Note: The figure plots scaled regression coefficients and 95 percent confidence intervals obtained from the estimation of equation (1) as reported in column (6) of Table 7. Point estimates and confidence intervals are scaled to reflect a 10 percent shock to both human and physical capital. The dependent variable is the total number of publications in department d and year t. The left hand panel reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies. The excluded year is 1931; the last data point before the dismissals. The right hand panel reports coefficients on the interaction of percentage destruction (by Allied bombings between 1941 and 1945) with year dummies. The excluded year is 1940; the last data point before the bombings.

The productivity of scientists follows a concave pattern over their lifetime (Levin and Stephan, 1991). Some of the changes in the age structure of departments are endogenous to the dismissals because departments with dismissals may have had to hire younger researchers, for example. In this case one should not control for these changes. Nonetheless, I investigate whether the results are driven by changes in the age structure of affected departments. I therefore adjust output by regressing individual productivity measures on a full set of age dummies (in 5 year bins). The residuals from this regression are then used to construct department productivity that I normalize to have a zero mean and a standard deviation of one. Using age-adjusted productivity as the dependent variable yields very similar results (Tables 5 and 6, column 7).

Different Samples

The main results were estimated with all German and Austrian universities that existed before 1926. Dropping Austrian science departments from the sample does not affect the results (Tables 5 and 6, column 8). The main sample includes all German universities that were based within the German borders of 1990, both in the FRG and the GDR. Reconstruction and re-hiring may have been different in the GDR compared to the market economy of the FRG. Dropping East German departments from the sample only slightly reduces the absolute magnitude of the coefficients suggesting that the results are not primarily driven by a different development in the GDR (Tables 5 and 6, column 9). These results also demonstrate that the special situation of the University of Berlin that was located in the Soviet sector does not drive

the results.³¹ The results are also robust to dropping both Austria and the GDR from the sample (Tables 5 and 6, column 10).

Different Shock Measures

As outlined above, most dismissals took place during the first years of the Nazi regime. Restricting the dismissals to these early dismissals (as in Waldinger, 2010) the results remain very similar (Tables 5 and 6, column 11).

The dismissal shock was confined to particular departments because the training of scientists is very subject specific. As a result, universities could not reallocate physicists to chemistry departments if a university had many dismissals in chemistry but few in physics. Following a bombing shock, however, universities could reallocate buildings across departments. This could have mitigated any negative effects of bombing destruction.

To investigate whether the reallocation of buildings is affecting the results, I measure destruction with average destruction in all science departments instead of destruction of individual science departments. Estimating the model with this alternative measure yields a more negative coefficient for 1950. The destruction of an additional 10 percent of science buildings lowered productivity by 0.11 instead of 0.05 standard deviations (Table 5, columns 6 and 12). In regressions that use citation weighted productivity as the dependent variable, the 1950 coefficient changes from -0.04 to -0.08 when I use average destruction in all science departments instead of the subject specific measure (Table 6, columns 6 and 12). These results suggest that universities could indeed mitigate some of the short-run effects of bombings by reallocating buildings across departments. When I measure the physical capital shock with average destruction across all science departments the negative effect persists until 1961 (significant at the 10 percent level) for the publication based productivity measure (Table 5, column 12) but not for the citation-weighted productivity measure (Table 6, column 12). By 1970, bombed departments had completely recovered independently of the destruction and productivity measures. In fact, citation-weighted productivity was significantly higher in bombed departments in 1970. This further suggests that bombed departments benefitted from upgrading during reconstruction.

Measurement Error in Bombing Destruction

The destruction measure of some departments may contain measurement error. To investigate how much measurement error attenuates the bombing results I instrument for department level destruction with a measure of bombing destruction at the university level.³² University level destruction was obtained from different data sources than the department level measure (see data appendix). As a result, the two measurement errors should be uncorrelated. The instrumental variable strategy should therefore minimize attenuation bias. First stage regressions are reported in Table A3 and indicate that university level destruction is a strong predictor of

³¹The University of Berlin was located in the Soviet sector of Berlin. It reopened in January 1946 (and was renamed into Humboldt University in 1949). In 1948, the Free University of Berlin was founded in the American sector of Berlin.

³²As the university level destruction measure captures destruction of all university buildings it is different from the average destruction in science departments that has been used above.

departments with 10 percent more destruction fell by 0.13 standard deviations in 1950. In 1961, departments with 10 percent more bombing destruction still had 0.11 standard deviations lower productivity. By 1970, however, productivity had recovered in bombed departments (Table 5, column 13). Equivalent results for citation-weighted publications indicate that productivity in departments with 10 percent more destruction fell by 0.08 standard deviations in 1950, and 0.05 standard deviations (significant at the 10 percent level) in 1961. By 1970, productivity had completely recovered (Table 6, column 13). These results suggest that measurement error may indeed attenuate the bombing results for 1950 and 1961. The finding that bombed departments had completely recovered by 1970, however, is not distorted by measurement error.

Interactions of Human and Physical Capital Shocks

The regression model estimated above does not allow for interactions of human and physical capital. To investigate whether such interactions are important I add triple interactions of # dismissed, % destruction, and year dummies to the regression. I.e. I estimate

Output_{dt} =
$$\beta_1 + \sum_t \beta_{2t} \text{HCShock}(1933-40)_d * \text{Year}_t + \sum_t \beta_{3t} \text{PCShock}(1942-45)_d * \text{Year}_t$$

+ $\sum_t \beta_{4t} \text{HCShock}(1933-40)_d * \text{PCShock}(1942-45)_d * \text{Year}_t$
+ $\beta_5 \text{DepartmentFE}_d + \beta_6 \text{YearFE}_t + \varepsilon_{dt}$ (2)

The first data point that could have been affected by both shocks is 1950. I therefore include the triple interactions for the years 1950, 1961, 1970, and 1980. The publication results show that departments with dismissals that were then bombed during WWII did significantly worse than other departments in 1950. The estimated coefficient indicates a reduction in publications by 0.03 standard deviations in departments that lost one scientist and then lost 10 percent of department buildings. As departments recovered quickly from the physical capital shock triple interactions with later years are no longer significant. (Table A4, column 2). For the citation weighted productivity measure all triple interactions are insignificant (Table A4, column 4). The dismissal results are remarkably robust to the inclusion of the triple interactions. The findings suggest that there are some complementarities between human and physical capital but that these are relatively minor. Furthermore, human capital effects are not driven by important complementarities with physical capital.

³³As I instrument for all interactions of year dummies with department level destruction, I estimate six first stage regressions: one for the interaction of % subject destruction with the 1926 dummy, one for the interaction of % subject destruction with the 1931 dummy, and so on. As a result, the usual F-test on the excluded instruments is not appropriate in this context. Stock and Yogo (2005) propose a test for weak instruments based on the Cragg-Donald (1993) Eigenvalue Statistic. Stock and Yogo (2005) only provide critical values for up to two endogenous regressors. With two endogenous regressors and two instruments the critical value is 7.03. Here, I use six instruments for six endogenous regressors. Appropriate critical values should be lower than 7.03. The Cragg-Donald EV statistic reported in Table A3 is 19.6. Weak instruments should therefore not bias the results.

4.3 Subject Specific Results

Data on the three subjects allow me to investigate whether the human and physical capital shocks had differential effects across disciplines. While physical capital may be more important in chemistry and some fields of physics (experimental physics, technical physics, or astrophysics) it is presumably less important in mathematics. The estimation results, however, indicate that the productivity effects of physical capital did not differ across fields (Table 7).

Despite the fact that most results are less precisely estimated because of smaller sample sizes, the dismissal results reveal interesting differences across subjects. If one considers citation weighted productivity, the results are largest and most persistent in mathematics, followed by physics (even though most coefficients in physics are not significant), and then chemistry (Table 7, columns 2, 4, and 6). These results suggests that high quality scientists are particularly important as dismissals in mathematics and physics were of higher quality than in chemistry (Table 1).

4.4 The Effect of High Quality Scientists

To further investigate this hypothesis I assign scientists to quality percentiles based on their pre-dismissal citation weighted publications. I then investigate how the dismissal of high quality scientists affected department productivity in the long run. Corresponding to the main specification reported above, the dismissal of one scientist lowered department productivity by between 0.18 and 0.32 standard deviations (Table 8, columns 1 and 2). The dismissal of an above median quality scientist lowered productivity by between 0.24 and 0.50 standard deviations (columns 3 and 4). The dismissal of higher quality scientists caused even larger productivity reductions that persisted in the long run. The dismissal of a scientist in the top 5th percentile led to a productivity reduction of 0.7 to 1.6 standard deviations (columns 9 and 10). Figure 7 summarizes these findings graphically.³⁴ The figure shows the relative productivity drop after the dismissal of a scientist of the relevant quality group. The higher the quality of dismissals the larger the long-run productivity decline.³⁵

³⁴To improve the clarity of the figure I do not report confidence intervals. The majority of estimated coefficients are significantly different from 0 (Table 8).

³⁵In regressions not reported here I investigate whether the dismissal results are entirely driven by very high quality scientists. For these test I exclude dismissed scientists whose citation-weighted publications place them in the top 5th percentile when I construct the dismissal variable. The effect of dismissals remains significant when I use this alternative dismissal measure.

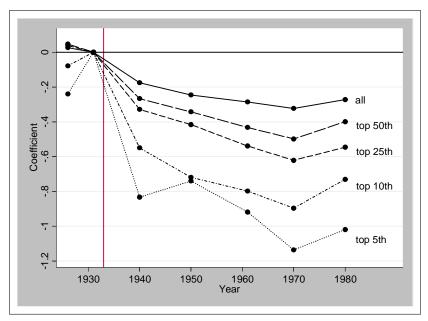


Figure 7: Persistence of high quality dismissals

Note: The figure plots regression coefficients reported in Table 8. The dependent variable is the total number of citation weighted publications in department d and year t. The top line reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies as in column (2) of Table 8. The second line from the top reports coefficients on the interaction of the number of dismissals of above median quality with year dummies as in column (4). The third line from the top reports coefficients on the interaction of the number of dismissals of scientists in the top 25th percentile with year dummies as in column (6), and so on. The excluded year is 1931. The regression also includes the interaction of percentage destruction with year dummies. To improve clarity, 95 percent confidence intervals are omitted from the graph. The large majority of regression coefficients is significantly different from 0 (Table 8).

5 Mechanisms for the Persistence of the Human Capital Shock

The results indicate that human capital is particularly important for the productivity of science departments. Furthermore, I find strong persistence of the human capital shock, with especially large declines after the dismissal of high quality scientists. In the following, I investigate possible mechanisms for the long-run persistence of the temporary dismissal shock. One possible explanation could be a fall department size from which departments never recovered. I investigate this hypothesis by regressing department size on the dismissal and destruction variables. Department size in departments with dismissals was significantly smaller until 1970 but by 1980 it had completely recovered (Table 9). These results indicate that a fall in department size can only explain some of the persistence of the human capital shock.

A related mechanism for the persistence of the human capital shock may be a permanent fall in the quality of hires. 'Star scientists' are often instrumental in attracting other high quality faculty. Before 1933, for example, the great mathematician David Hilbert was instrumental in attracting theoretical physicist Max Born (Nobel Prize, 1954) to the University of Göttingen. Max Born then used his influence to hire experimental physicist James Franck (Nobel

Prize, 1925) (Jungk, 1963 pp. 22-23). Born and Franck were dismissed from the University of Göttingen in 1933. It is quite likely that these and other dismissals had permanent effects on the quality of subsequent hires. I therefore investigate whether hiring quality declined in departments with dismissals. I also analyze whether the dismissal of high quality scientists caused particularly large reductions in the quality of new hires. To identify hires I need to observe changes in the composition of departments. As a result, I can measure the quality of hires from 1931 onwards. For 1931, I classify all scientists who joined a department between 1926 and 1931 as a hire and measure the average quality of these hires. I construct equivalent measures for the quality of hires in subsequent years.

I regress average hiring quality on the number of dismissals interacted with year indicators. The dismissal of one scientist lowered hiring quality by between 0.05 and 0.06 standard deviations. Estimated effects are significant for all years and persist until 1980 (Table 10, column 1). Additional results show that hiring quality dropped more after losing high quality scientists, in particular after losing scientists of exceptional quality (columns 2 to 5). Figure 8 shows the reduction in hiring quality after the dismissal of scientists with different qualities.

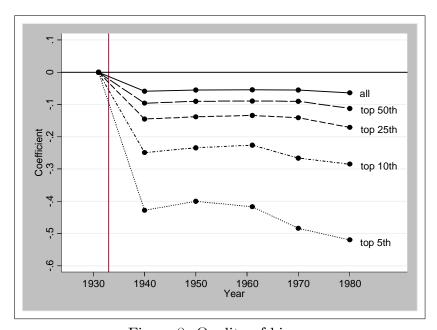


Figure 8: Quality of hires

Note: The figure plots regression coefficients as reported in Table 10. The dependent variable is quality of new hires in department d and year t. The top line reports coefficients on the interaction of the number of dismissals (between 1933 and 1940) with year dummies as in column (1) of Table 10. The second line from the top reports coefficients on the interaction of the number of dismissals of above median quality with year dummies as in column (2) and so on. The excluded year is 1931. The regression also includes the interaction of percentage destruction with year dummies and all control variables. To improve clarity, 95 percent confidence intervals are omitted from the graph. All regression coefficients are significantly different from 0 (Table 10).

The fall in hiring quality was partly driven by a decline in the quality of PhD students that were produced in departments with dismissals after 1933 (Waldinger, 2010). The decline in faculty quality was perpetuated because German universities often hire former PhD students.³⁷

³⁶The quality of hires is measured by career citation weighted publications which I normalize to have a zero mean and a standard deviation of one within subjects.

³⁷ Almost 20 percent of mathematics PhD students who graduated between 1912 and 1940 and who obtained a

Another potential mechanism for the persistence of the human shock could be localized productivity spillovers. The productivity of scientists who remained in departments with dismissals after 1933 could have declined because of a reduction in the number and quality of local peers. The potential productivity decline for scientists who were present in 1933 could have affected later generations of scientists in departments with dismissals. In this case, however, localized spillovers are unlikely to drive persistence because even the productivity of scientists who directly experienced the dismissals did not decline. Not even the loss of high quality colleagues had negative productivity effects on contemporaneous local peers (Waldinger, 2012).

6 Conclusion

I use the dismissal of scientists in Nazi Germany and Allied bombings during WWII as exogenous and temporary shocks to the human and physical capital of science departments. While I find evidence that both shocks had negative productivity effects in the short run, departments that suffered from physical destruction recovered by 1961. By 1970, departments with buildings that were destroyed during WWII even did slightly better than departments without bombing destruction. The human capital shock, however, had a larger negative effect on productivity in the short run and was much more persistent. I also show that the dismissal of high quality scientists caused particularly large productivity reductions. The findings indicate that human capital is potentially more important than physical capital for the productivity of science departments.

My findings, of course, do not indicate that physical capital is irrelevant for productivity because post-war reconstruction targeted destroyed departments. It seems, however, that negative shocks to physical capital can be overcome much more easily than human capital shocks. In recent years, some fields of science have become more dependent on large capital expenditures such as particle accelerators. Negative physical shocks in these fields may lead to more persistent effects on productivity than shocks that have occurred in the past.

The persistence of the human capital shock is particularly remarkable if one considers that most scientists who had been employed by the former German universities (Breslau now Wroclaw, TU Breslau, and Königsberg now Kaliningrad) that became part of Poland and the Soviet Union had to relocate to universities that remained on German territory after 1945. Departments that had lost people during the dismissals could have hired the best people from these universities.

I show that an important mechanism for the persistence of the human capital shock was a permanent fall in the quality of new hires, in particular after the dismissal of high quality scientists. 'Star scientists' seem to be especially valuable as they attract other high quality researchers. This suggests that appointing very high quality scholars may be a good strategy if

university position in Germany returned to the university that had granted their PhD. The Pearson chi-squared test statistic to test the hypothesis of independence of the PhD university and the university of employment is 1,300. The hypothesis is thus rejected with a p-value smaller than 0.001.

a university of country wanted to raise its research profile. In fact, anecdotal evidence suggests that hiring high-quality faculty during the 1980s and 1990s has been instrumental for the recent rise of New York University and in particular their economics department (Honan, 1995).

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7 Tables

Table 1: Number and Quality of Dismissed Scientists

			Physics			C	Chemistry			Ma	Mathematics	
				% of citation				% of citation				% of citation
				weighted				weighted				weighted
		% of all	% of pubs.	pubs.		% of all	% of pubs.	pubs.		% of all	% of pubs.	pubs.
Year of	# of	Scientists	published	published	# of	Scientists	published	published	# of	Scientists	published	published
Dismissal	Dismissals	in 1931	by dismissed	by dismissed	Dismissals	in 1931	by dismissed	by dismissed	Dismissals	in 1931	by dismissed	by dismissed
1933	32	9 4	13.6	50.3	88	×	11.2	11.9	2.6	11.7	23.6	
1934	3 9	: «) x	1.4	} oc	9:1		: F	; ¬	1.7	?; « «	0.0
1935	ာဏ	o. ∞	1.6	: L	ော	1.3	2.7	. r.	, rd	2.2	ა დ ე ი)
1936	0	0.0	0.0	0.0		0.2	. i 0	9:0	0	0.0); c	0.0
1937		0.3	0.3	0.0	က	9.0	0.3	0.5	2	6.0	1.7	5.7
1938	က	0.0	9.0	0.1	6	1.8	5.5	7.3	က	1.3	0.0	3.3
1939	2	9.0	9.0	0.0	က	9.0	0.8	0.7	1	0.4	0.3	0.0
1940	1	0.3	0.2	1.6	1	0.2	0.1	0.1	1	0.4	0.3	0.2
1933-40	51	15.0	23.8	64.0	69	14.1	22.0	23.4	43	18.7	31.0	61.3
						Ш.		7007	,			

of Dismissals is the total number of dismissals in a subject (physics, chemistry, or mathematics) in a given year. % of all Scientists in 1931 reports dismissals as percentage of total faculty in 1931. % of pubs. published by dismissed reports the percentage of publications that were published by the dismissed between 1928 and 1932. % of citation weighted pubs. published by the dismissed between 1932.

Table 2: Dismissal and Bombing Shocks Across Science Departments

		I	Physics			Chemistry			M	Mathematics			
				Bombing			Bombing				Bombing		
	Dismissal Shock	al Sh	ock	Shock	Dismiss	Dismissal Shock	Shock	Dismissal Shock	al She	ck	Shock	Total Destruction	truction
	Jo #	Dis	Dismissed	Destruct.	# of	Dismissed	Destruct.	# of	Disı	Dismissed	Destruction	University	City
	scienstists	က	33-40	40-45	scienstists	33-40	40-45	scienstists	60	33-40	40-45	Destruct.	Destruct.
University	(1931)	#	in %	in %	(1931)	% ui #	% ui	(1931)	#	in %	% ui	in %	in %
Aachen TU	5		20.0	20.4	11		52.4	9	2	33.3	25.0	20	49
Berlin	41	10	24.4	10.0	47	16 34.0	65.0	14	5	35.7	10.0	45.8	37
Berlin TU	30	6	30.0	25.0	41	11 26.8	11.1	17	5	29.4	48.0	48	37
Bonn	10	П	10.0	50.0	14	2 14.3	20.6	∞	1	12.5	20.6	40	24
Braunschweig TU	2	0	0	90.0	11	$\theta = 0$	47.0	2	0	0	25	7.0	26
Darmstadt TU	10	က	30.0	m	12	4 33.3	m	2	1	20.0	m	7.5	46
Dresden TU	11	П	9.1	100.0	17	1 5.9	5.0	∞	0	0	100.0	92	39
Erlangen	2	0	0	0	6	1 11.1	0	က	0	0	0	0	4.8
Frankfurt	13	2	15.4	37.0	18	5 27.8	57.0	∞	4	50.0	27.0	09	32
Freiburg	2	_	20.0	100.0	11	2 18.2	0.09	2	1	20.0	85.0	72.5	28
Giessen	9	П	16.7	50.0	6	$\theta = 0$	100.0	4	0	0	50.0	67.5	53
Göttingen	20	∞	40.0	0	17	3 17.6	0	16	10	62.5	0	1.7	2.1
Graz	7	П	14.3	10.0	∞	0 0	0	9	0	0	0	5	33
Graz TU	1	0	0	0	2	0 0	0	20	0	0	50.0	20	33
Greifswald	7	0	0	0	4	0 0	0	4	0	0	0	0	0
Halle	4	0	0	0	2	1 14.3	0	2	П	20.0	0	5	20
Hamburg	15	2	13.3	30.0	12	2 16.7	30.0	∞	1	12.5	15.0	20	54
Hannover TU	4	0	0	22.2	10	0 0	37.5	4	0	0	22.2	41.3	47
Heidelberg	9	0	0	0	19	2 10.5	0	2	3	0.09	0	0	1
$_{\rm Innsbruck}$	9	0	0	0	∞	0 0	50.0	5	0	0	0	ш	09
Jena	14	П	7.1	0	10	0 0	62.5	ъ	0	0	50.0	87.3	20
Karlsruhe TU	2	\vdash	20.0	75.0	16	5 31.3	100.0	20	1	20.0	75.0	70	26
Kiel	2	\vdash	14.3	62.5	∞	0 0	50.0	2	2	40.0	75.0	09	41
Köln	9	\vdash	16.7	2.99	9	0 0	50.0	2	П	80.0	0	20	44
Leipzig	12	2	16.7	41.0	21	2 9.5	100.0	∞	2	25.0	0	70	19
Marburg	J.	0	0	0	∞	0 0	50.0	9	0	0	0	16.3	4
München	11	2	18.2	42.0	19	3 15.8	95.0	∞	П	12.5	70.0	70	32
München TU	15	П	6.7	36.7	17	1 5.9	30.0	20	0	0	50.0	80	32
Münster	4	0	0	100.0	7	0.0	100.0	9	0	0	75.0	75.3	49
${ m Rostock}$	4	0	0	0	9	0.0	0	1	0	0	0	0	40
Stuttgart TU	7	П	14.3	76.7	10	1 10.0	2.99	6	0	0	40	80	35
Tübingen	က	0	0	0	6	0.0	0	4	0	0	0	0	5
Wien	4	0	0	25.0	30	5 16.7	37.5	6	2	22.2	25.0	30	28
Wien TU	∞	0	0	0	19	1 5.3	40.0	13	0	0	0	13.3	28
Würzburg	29	က	10.3	90.0	12	1 8.3	0.06	က	0	0	0.06	82.5	7.5
(1001)		-			5, 1001								

of scientists (1931) reports the number of scientists in a department in 1931 (before the dismissal shock). Dismissed 33-40 reports the number of dismissals in a department between 1940. The total number of dismissal is slightly higher than in Table 1 because professors with appointments in two universities are reported twice in this table. Destruct. 40-45 reports percentage destruction of department facilities due to Allied bombings between 1940 and 1945. University Destruct. reports the percentages of university facilities that were destruction, and city level destruction and city level destruction data come from Hohn (1994) and other sources. See data appendix for details.

- Publications
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Table

	(1)	(6)	(3)	(V)	(5)	(9)	(2)	(8)	(6)	(10)
	(I)	(2)	(6)	(±)	(6)	(0)	(1)	(0)	(6)	(10)
	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-
Dependent Variable:	cations	cations	cations	cations	cations	cations	cations	cations	cations	cations
Number of Dismissals * 1926	0.023	0.017	0.017				0.024	0.018	0.018	0.018
	(0.025)	(0.025)	(0.025)				(0.025)	(0.026)	(0.026)	(0.026)
Number of Dismissals $*$ 1940	-0.173***	-0.173***	-0.173***				-0.172***	-0.171***	-0.171***	-0.170***
	(0.037)	(0.039)	(0.039)				(0.039)	(0.041)	(0.041)	(0.042)
Number of Dismissals $*$ 1950	-0.210*	-0.222*	-0.219**				-0.210	-0.221*	-0.218**	-0.219**
	(0.121)	(0.125)	(0.099)				(0.124)	(0.128)	(0.102)	(0.102)
Number of Dismissals * 1961	-0.245**	-0.257**	-0.254***				-0.244**	-0.255**	-0.253***	-0.253***
	(0.102)	(0.105)	(0.080)				(0.105)	(0.107)	(0.083)	(0.084)
Number of Dismissals * 1970	-0.291***	-0.286**	-0.283***				-0.290**	-0.283**	-0.281***	-0.282***
	(0.104)	(0.108)	(0.084)				(0.108)	(0.111)	(0.087)	(0.088)
Number of Dismissals * 1980	-0.202**	-0.202**	-0.207***				-0.199**	-0.198**	-0.202***	-0.212***
	(0.077)	(0.070)	(0.071)				(0.077)	(0.078)	(0.070)	(0.077)
% Destruction * 1926				-0.004	-0.005*	-0.005*	-0.004*	-0.004**	-0.004**	-0.004*
				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1931				-0.008***	-0.008***	-0.008***	***800.0-	-0.008***	***800.0-	-0.007***
				(0.002)	(0.002)	(0.002)	(0.001)	(0.002)	(0.002)	(0.002)
% Destruction * 1950				-0.002	-0.003*	-0.005**	-0.003	-0.003*	-0.005**	-0.004**
				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1961				0.000	-0.001	-0.003	0.000	-0.001	-0.003	-0.002
				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1970				0.003*	0.004*	0.002	0.003	0.004*	0.002	0.002**
÷				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1980				(0.003)	(0.001)	-0.000 (0.004)	0.001 (0.003)	(0.001)	(0.004)	(0.004)
Department FE	SOA	ves	SOA	Ves	ves	Ves	ves	Ves	Ves	Ves
Year FE	ves			ves	,	•	ves			,
Subject*Year FE	•	ves	ves	•	ves	ves	•	ves	ves	ves
Occupation Zone * Post45		,	ves			ves		•	ves	ves
% City Destruction * Year Dummies	mies		5			GO C			a c	yes
Observations	714	714	714	714	714	714	714	714	714	714
R-squared	0.590	0.650	0.678	0.513	0.576	0.603	0.603	0.664	0.689	0.693
***significant at 1% level	**significant at 5% level	level	*significant at 10% level	% level	(All standard	errors clustered a	(All standard errors clustered at university level			

The dependent variable Publications is the sum of publications published by all scientists in department d in a five-year window around year t. The variable is normalized to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * 1926 is equal The excluded interaction is % destruction with 1940, the last observation before the bombings. Department FE is the full set of 105 department fixed effects. Year FE is a set of year dummies for each year 1931, 1940, 1950, 1961, 1970, and 1980 (1926 is the excluded year). Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. (All standard errors clustered at university level) *significant at 10% level Year Dummies is the interaction of city level destruction with the full set of year dummies. **significant at 5% level **significant at 1% level

Weighted Publications
Citation \
Shocks -
d and Bombing
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e of
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4
Table

	(1)	(2)	(3)	(4)	(2)	(9)	(7)	(<u>&</u>	(6)	(10)
	Citation	Citation	Citation	Citation	Citation	Citation	Citation	Citation	Citation	Citation
	weighted	weighted	weighted	weighted	weighted	weighted	weighted	weighted	weighted	weighted
Dependent Variable:	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.
Number of Dismissals * 1926	-0.093	-0.100	-0.100				-0.092	-0.099	-0.099	-0.099
	(0.076)	(0.078)	(0.078)				(0.076)	(0.078)	(0.078)	(0.070)
Number of Dismissals * 1940	-0.181***	-0.188***	-0.188***				-0.181***	-0.186***	-0.186***	-0.185***
	(0.063)	(0.062)	(0.062)				(0.060)	(0.059)	(0.059)	(0.059)
Number of Dismissals $*$ 1950	-0.195***	-0.201***	-0.196***				-0.195***	-0.200***	-0.196***	-0.197***
	(0.030)	(0.036)	(0.031)				(0.031)	(0.036)	(0.032)	(0.032)
Number of Dismissals $*$ 1961	-0.190***	-0.207***	-0.202***				-0.189***	-0.206***	-0.202***	-0.205***
Number of Dismissals * 1970	(0.036) $-0.210***$	(0.041) $-0.220***$	(0.042) $-0.215***$				(0.034) $-0.209***$	(0.040) $-0.217***$	(0.042) $-0.212***$	(0.042) $-0.215***$
	(0.047)	(0.049)	(0.053)				(0.038)	(0.042)	(0.046)	(0.049)
Number of Dismissals $*$ 1980	-0.129*	-0.142**	-0.141**				-0.126*	-0.139**	-0.138**	-0.158**
	(0.065)	(0.059)	(0.061)				(0.059)	(0.055)	(0.057)	(0.064)
% Destruction * 1926				-0.003	-0.003	-0.003	-0.003	-0.003	-0.003	-0.002
				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1931				**900.0-	*900.0-	*900.0-	**900.0-	-0.005**	-0.005**	-0.005
				(0.003)	(0.003)	(0.003)	(0.003)	(0.002)	(0.002)	(0.003)
% Destruction * 1950				-0.002	-0.002	-0.004*	-0.002	-0.002	-0.004*	-0.002
				(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1961				-0.001	-0.001	-0.003	-0.001	-0.001	-0.003	0.000
				(0.003)	(0.003)	(0.003)	(0.003)	(0.003)	(0.003)	(0.003)
$\%$ Destruction * 1970				0.006*	0.006*	0.005	0.006* (0.003)	0.006*	0.005	0.008*
% Destruction * 1980				0.001	0.001	-0.000	0.001	0.001	-0.000	0.006
				(0.005)	(0.005)	(0.005)	(0.005)	(0.005)	(0.005)	(0.005)
Department FE	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Year FE	yes			yes			yes			
Subject*Year FE		yes	yes		yes	yes		yes	yes	yes
Occupation Zone * Post1945 Dummies			yes			yes			yes	yes
City Level Destruction * Year Dummies										yes
Observations	714	714	714	714	714	714	714	714	714	714
R-squared	0.460	0.484	0.496	0.440	0.460	0.471	0.475	0.497	0.507	0.518

to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is % Destruction * 1926 is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is % destruction with 1940, the last observation before the bombings. Department FE is the full set of 105 department fixed effects. Year FE is a set of year dummies for each year 1931, 1940, 1950, 1961, 1970, and 1980 (1926 is the excluded year). Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a The dependent variable Citation weighted Pubs. is the sum of citation-weighted publications published by all scientists in department d in a five-year window around year t. The variable is normalized equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. (All standard errors clustered at university level) post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. *significant at 10% level **significant at 5% level ***significant at 1% level

	(1)	(2)	(3)	(4)	(2)	(9)	(7)	8	(6)	(10)	(11)	(12)	(13)
	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-	Age-adj.	Publi-	Publi-	Publi-	Publi-	Publi-	Publi-
Dependent Variable:	cations	cations	cations	cations	cations	cations	Pubs.	cations	cations	cations	cations	cations	cations
	Full	Full	Full	Full	Full	Full	Full	Dropping	Dropping	Dropping	Full	Full	Full
Sample:	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	Austria	East Ger.	$_{ m both}$	$_{ m Sample}$	$_{ m Sample}$	Sample
											only 33-34 dismissals	Science Dest.	Instrumenting with Uni. Dest.
# of Dismissals * 1926	0.018	0.020	0.020	0.026	0.026	0.026	0.086**	0.024	0.027	0.020	0.048	0.028	0.020
	(0.026)	(0.026)	(0.026)	(0.026)	(0.026)	(0.026)	(0.036)	(0.029)	(0.043)	(0.041)	(0.041)	(0.026)	(0.023)
# of Dismissals * 1940	-0.170***	-0.173***	-0.173***	-0.165***	-0.175***	-0.175***	-0.160***	-0.138***	-0.116***	-0.100**	-0.153***	-0.175***	-0.183***
	(0.042)	(0.042)	(0.042)	(0.037)	(0.038)	(0.038)	(0.053)	(0.041)	(0.033)	(0.037)	(0.040)	(0.036)	(0.038)
# of Dismissals * 1950	-0.236**	-0.244**	-0.246**	-0.236**	-0.248**	-0.246**	-0.178	-0.221***	-0.080	-0.045	-0.209*	-0.242**	-0.255***
	(0.102)	(0.103)	(0.103)	(0.102)	(0.103)	(0.099)	(0.122)	(0.070)	(0.047)	(0.042)	(0.114)	(0.098)	(0.091)
# of Dismissals * 1961	-0.271***	-0.282***	-0.284***	-0.276***	-0.288***	-0.286***	-0.233**	-0.259***	-0.151***	-0.114**	-0.275***	-0.283***	-0.292***
	(0.078)	(0.077)	(0.075)	(0.075)	(0.075)	(0.071)	(0.090)	(0.051)	(0.048)	(0.050)	(0.081)	(0.069)	(0.066)
# of Dismissals * 1970	-0.300***	-0.314***	-0.316***	-0.314***	-0.326***	-0.323***	-0.285***	-0.298***	-0.203***	-0.169**	-0.314**	-0.322***	-0.324***
	(0.079)	(0.078)	(0.073)	(0.074)	(0.073)	(0.069)	(0.074)	(0.050)	(0.069)	(0.073)	(0.081)	(0.066)	(0.063)
# of Dismissals * 1980	-0.234***	-0.251***	-0.258***	-0.259***	-0.272***	-0.272***	-0.287***	-0.271***	-0.184***	-0.162**	-0.238***	-0.271***	-0.267***
:	(0.067)	(0.065)	(0.063)	(0.063)	(0.062)	(0.061)	(0.087)	(0.053)	(0.065)	(0.065)	(0.059)	(0.061)	(0.053)
% Destruction * 1926	-0.004*	-0.004*	-0.004*	-0.003	-0.002	-0.002	-0.004	-0.004	-0.003	-0.006	-0.002	-0.003	***800.0-
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.002)	(0.004)	(0.005)	(0.003)	(0.004)	(0.003)
% Destruction * 1931	-0.007**	***200.0-	-0.007***	-0.005**	-0.005*	-0.005*	*900.0-	***900.0-	-0.005*	-0.008**	-0.004	-0.003	-0.011***
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.002)	(0.003)	(0.003)	(0.003)	(0.003)	(0.003)
% Destruction * 1950	-0.006***	-0.005***	***900.0-	-0.005**	-0.005**	-0.005**	-0.005	-0.008**	-0.007*	+600.0-	**900.0-	-0.011**	-0.013***
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.003)	(0.004)	(0.005)	(0.002)	(0.004)	(0.003)
% Destruction * 1961	-0.004**	-0.003*	-0.004**	-0.003	-0.003	-0.003	-0.004	-0.005***	-0.002	-0.003	-0.004*	+900.0-	-0.011***
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.002)	(0.003)	(0.004)	(0.002)	(0.003)	(0.003)
% Destruction * 1970	0.003	0.003	0.003	0.004	0.004	0.003	0.000	0.001	0.004	0.003	0.002	0.004	-0.002
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.002)	(0.004)	(0.005)	(0.002)	(0.004)	(0.003)
% Destruction * 1980	0.001	0.002	0.001	0.001	0.001	0.001	-0.000	-0.003	0.001	-0.001	0.001	0.001	-0.008
	(0.004)	(0.004)	(0.003)	(0.003)	(0.003)	(0.003)	(0.004)	(0.005)	(0.005)	(0.006)	(0.004)	(0.004)	(0.006)
Standard Controls	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Länder Dummies * P45	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Quadratic in Uni. Age		yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
# of Deps. Within 50km			yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Industries (1933) * Year				yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Fract. Jews (1933) * P45					yes	yes	yes	yes	yes	yes	yes	yes	yes
Dist. to Iron Curtain * P45						yes	yes	yes	yes	yes	yes	yes	yes
Observations	714	714	714	714	714	714	714	609	288	483	714	714	714
R-squared	0.703	0.706	0.708	0.716	0.717	0.718	0.603	0.724	0.698	0.685	0.704	0.719	
***significant at 1%	**significant at 5%	ut 5%	*significant at 10%	at 10%	(s.e. clu	stered at uni	(s.e. clustered at university level)						

***significant at 1% **significant at 10% (s.e. clustered at university level)

The dependent variable Publications is the sum of publications published by all scientists in department d in a five-year window around year t. The variable is normalized to have zero mean and a standard deviation of one within subjects. In column (7) the dependent variable is equal is age adjusted publications, also normalized to have zero mean and a standard deviation of one within

subjects. # of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction ** 1926 is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is destruction with 1940, the last observation before the bombings. Standard Controls are all controls as reported in column (10) of Table 3, i.e. Department FEs, Subject*Year FEs, Occupation Quadratic in Uni. Age is equal to the age of the university in each year and its square. # of Deps. Within 50km measures the number of departments with the same subject within 50 kilometers of each department in each year. Industries (1933) * Year is the fraction of firms in a city belonging to each of 3 armament related universities in 1933 (1930 for Austria) interacted with a full set of year dummies. The 3 industries are: iron and steel production, mechanical engineering and vehicle construction, and chemical industry. Fract. Jews (1933) * P45 is the fraction of Jews in each Zones * Post45, and % City Destruction * Year Dummies. Länder Dummies. * P45 is a set of dummy variables for each post-war German federal state (Land) interacted with a post-1945 dummy. city in 1933 interacted with a post 1933 dummy. Dist. to Iron Curtain * P45 is the distance to the iron curtain from each city interacted with a post-1945 dummy.

Table 6: Robustness Checks - Citation weighted Publications

	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)	(11)	(12)	(13)
	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.	Age-adj. Cit.	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.	Cit. weig.
Dependent Variable:	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	weig. Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.	Pubs.
	Full	Full	Full	Full	Full	Full	Full	Dropping	Dropping	Dropping	Full	Full	Full
Sample:	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	$_{ m Sample}$	Sample	Austria	East Ger.	$_{ m both}$	Sample	$_{ m Sample}$	Sample
											only 33-34	Science	Instrument.
											dismissals	Dest.	w/ Uni. Dest.
# of Dismissals * 1926	660.0-	860.0-	660:0-	-0.087	-0.087	-0.087	920.0-	-0.080	-0.154*	-0.156*	-0.080	-0.084	-0.099
	(0.070)	(0.070)	(0.070)	(0.077)	(0.077)	(0.077)	(0.098)	(0.078)	(0.090)	(0.083)	(0.089)	(0.078)	(0.069)
# of Dismissals * 1940	-0.185***	-0.186***	-0.185***	-0.180***	-0.180***	-0.181***	-0.188**	-0.150*	-0.248***	-0.254***	-0.177**	-0.181**	-0.190***
	(0.059)	(0.060)	(0.060)	(0.062)	(0.064)	(0.064)	(0.076)	(0.087)	(0.057)	(0.056)	(0.074)	(0.067)	(0.054)
# of Dismissals * 1950	-0.188***	-0.190***	-0.195***	-0.191***	-0.191***	-0.187***	-0.150***	-0.156***	-0.177***	-0.177***	-0.173***	-0.182***	-0.196***
	(0.031)	(0.031)	(0.042)	(0.031)	(0.027)	(0.028)	(0.030)	(0.040)	(0.048)	(0.053)	(0.031)	(0.029)	(0.035)
# of Dismissals * 1961	-0.196***	-0.198***	-0.202***	-0.196***	-0.196***	-0.191***	-0.157***	-0.163***	-0.196***	-0.194***	-0.192***	-0.188***	-0.203***
	(0.038)	(0.039)	(0.046)	(0.038)	(0.037)	(0.039)	(0.047)	(0.054)	(0.062)	(0.067)	(0.048)	(0.041)	(0.039)
# of Dismissals * 1970	-0.206***	-0.209***	-0.213***	-0.220***	-0.220***	-0.215***	-0.194***	-0.186***	-0.221**	-0.211**	-0.229***	-0.212***	-0.214***
	(0.046)	(0.046)	(0.049)	(0.046)	(0.048)	(0.050)	(0.060)	(0.065)	(0.083)	(0.086)	(0.063)	(0.055)	(0.044)
# of Dismissals * 1980	-0.146**	-0.150**	-0.163**	-0.181***	-0.181***	-0.181***	-0.191***	-0.168**	-0.197**	-0.207**	-0.161*	-0.178**	-0.165**
	(0.066)	(0.066)	(0.072)	(0.061)	(0.064)	(0.064)	(0.067)	(0.076)	(0.086)	(0.088)	(0.088)	(0.066)	(0.066)
0/ Dest. 21006	600 0	6000	6000	6000	6000	6000	000	700 0	0 001	600 0	6000	7000	6000
% Destruction 1920	-0.002	-0.002	-0.002	-0.00	enn.n-	-0.00	-0.004	-0.004	-0.001	-0.002	-0.00	-0.004	-0.00
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.002)	(0.002)	(0.003)	(0.002)	(0.003)	(0.002)
% Destruction * 1931	-0.005	-0.005	-0.005	-0.003	-0.003	-0.003	-0.004	-0.003	0.003	0.004	-0.002	-0.001	-0.007**
	(0.003)	(0.003)	(0.003)	(0.003)	(0.004)	(0.004)	(0.005)	(0.004)	(0.004)	(0.005)	(0.004)	(0.003)	(0.003)
% Destruction * 1950	-0.003	-0.003	-0.003*	-0.004*	-0.004*	-0.004*	-0.005	-0.005*	-0.002	-0.003	-0.004*	-0.008**	***800.0-
	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	(0.003)	(0.003)	(0.003)	(0.004)	(0.002)	(0.003)	(0.003)
% Destruction * 1961	-0.001	-0.001	-0.001	-0.002	-0.002	-0.002	-0.003	-0.003	0.001	0.001	-0.002	-0.004	-0.005*
	(0.003)	(0.003)	(0.003)	(0.003)	(0.003)	(0.003)	(0.004)	(0.003)	(0.005)	(0.005)	(0.003)	(0.003)	(0.003)
% Destruction * 1970	0.006	0.006	0.006	0.007	0.007	0.007	0.006	0.007	0.013**	0.016*	0.007	0.009**	0.004
	(0.004)	(0.004)	(0.004)	(0.004)	(0.004)	(0.004)	(0.005)	(0.005)	(0.000)	(0.008)	(0.004)	(0.004)	(0.003)
% Destruction * 1980	0.005	0.005	0.004	0.009	0.009	0.009	0.010	0.009	0.012*	0.013	0.009	0.004	-0.009
	(0.005)	(0.005)	(0.005)	(0.007)	(0.007)	(0.007)	(0.008)	(0.000)	(0.007)	(0.009)	(0.007)	(0.005)	(0.007)
Standard Controls	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Länder Dummies * P45	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Quadratic in Uni. Age		yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
# of Deps. within 50km			yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Industries (1933) * Year				yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Fract. Jews (1933) * P45					yes	yes	yes	yes	yes	yes	yes	yes	yes
Dist. to Iron Curt. * P45						yes	yes	yes	yes	yes	yes	yes	yes
Observations	714	714	714	714	714	714	714	609	288	483	714	714	714
R-squared	0.525	0.525	0.526	0.528	0.528	0.552	0.367	0.558	0.571	0.579	0.548	0.548	0.517
****	**sionificant at 5%	at 5%	*significant at 10%	t at 10%	(s.e. clu	(s.e. clustered at university level)	versity level)						

***significant at 1% **significant at 5% *significant at 10% (s.e. clustered at university level)

The dependent variable Cit. weig. Pubs. is the sum of citation weighted publications published by all scientists in department d in a five-year window around year t. The variable is normalized to have a zero mean and a standard deviation of one within subjects. In column (7) the dependent variable is equal is age adjusted citation weighted publications, also normalized to have zero mean and a standard deviation of one. All explanatory variables are defined as in Table 5.

Table 7: Individual Subjects

	(1)	(2)	(3)	(4)	(5)	(6)
		Citation		Citation		Citation
	Publi-	weighted	Publi-	weighted	Publi-	weighted
Dependent Variable:	cations	Pubs.	cations	Pubs.	cations	Pubs.
	Phy	rsics	Chen	nistry	Math	ematics
Number of Dismissals * 1926	0.164	-0.033	0.026	0.012	-0.180	-0.469**
	(0.140)	(0.286)	(0.016)	(0.038)	(0.181)	(0.187)
Number of Dismissals * 1940	-0.112**	-0.181	-0.178***	-0.053**	-0.295*	-0.553***
	(0.044)	(0.227)	(0.060)	(0.020)	(0.162)	(0.192)
Number of Dismissals * 1950	-0.059	-0.122	-0.317***	-0.101*	-0.445**	-0.606***
	(0.190)	(0.248)	(0.110)	(0.054)	(0.182)	(0.182)
Number of Dismissals * 1961	-0.215**	-0.277	-0.309***	-0.060	-0.440**	-0.472**
	(0.089)	(0.189)	(0.089)	(0.084)	(0.178)	(0.202)
Number of Dismissals * 1970	-0.271***	-0.314	-0.346***	-0.101	-0.454**	-0.500**
	(0.080)	(0.204)	(0.096)	(0.061)	(0.177)	(0.200)
Number of Dismissals * 1980	-0.291**	-0.436**	-0.247***	0.078	-0.404**	-0.475**
	(0.119)	(0.186)	(0.089)	(0.082)	(0.165)	(0.175)
% Destruction * 1926	-0.002	-0.001	0.005	0.003	-0.007	-0.007
,,	(0.005)	(0.002)	(0.005)	(0.003)	(0.005)	(0.006)
% Destruction * 1931	-0.002	-0.000	0.001	-0.001	-0.013**	-0.005
, ,	(0.003)	(0.002)	(0.007)	(0.004)	(0.006)	(0.009)
% Destruction * 1950	0.000	-0.001	0.000	0.002	-0.007	-0.004
,,	(0.005)	(0.004)	(0.007)	(0.006)	(0.005)	(0.006)
% Destruction * 1961	0.001	-0.002	0.003	0.005	-0.007	-0.005
,,	(0.003)	(0.003)	(0.008)	(0.010)	(0.006)	(0.006)
% Destruction * 1970	0.003	0.001	0.008	0.018	0.008	0.006
,,	(0.004)	(0.003)	(0.007)	(0.012)	(0.005)	(0.007)
% Destruction * 1980	-0.002	0.002	0.017*	0.023**	-0.003	0.002
	(0.007)	(0.008)	(0.009)	(0.011)	(0.007)	(0.008)
Department FE	yes	yes	yes	yes	yes	yes
Subject*Year FE	yes	yes	yes	yes	yes	yes
Occupation Zones * Post45	yes	yes	yes	yes	yes	yes
% City Destruction * Year Dummies	yes	yes	yes	yes	yes	yes
All additional controls	yes	yes	yes	yes	yes	yes
Observations	238	238	238	238	238	238
R-squared	0.761	0.645	0.868	0.761	0.750	0.666
**significant at 1% **significant		*significant			ered at univer	

The dependent variable Publications reported in odd columns is the sum of publications published by all scientists in department d in a five-year window around year t. The dependent variable Citation weighted Pubs. reported in even columns is the sum of publications published by all scientists in department d in a five-year window around year t. Dependent variables are normalized to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals in Nazi Germany between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * 1926 is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is % destruction with 1940, the last observation before the bombings. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6.

Table 8: Dismissal of Top Scientists

	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)
		Citation		Citation		Citation		Citation		Citation
	Publi-	weighted	Publi-	weighted	Publi-	weighted	Publi-	weighted	Publi-	weighted
Dependent Variable:	cations	Pubs.	cations	Pubs.	cations	Pubs.	cations	Pubs.	cations	Pubs.
	All Dis	All Dismissals	Above Mec	Above Median Quality	Top 25th	Top 25th percentile	Top 10th percentile	percentile	Top 5th percentile	ercentile
Number of Dismissals * 1926	0.026	-0.087	0.047*	-0.098	0.039	-0.161	-0.079	-0.471**	-0.241	-0.784*
	(0.026)	(0.077)	(0.028)	(0.071)	(0.057)	(0.127)	(0.067)	(0.202)	(0.237)	(0.444)
Number of Dismissals * 1940	-0.175***	-0.181***	-0.267***	-0.243***	-0.329***	-0.349***	-0.550***	-0.712***	-0.833***	-1.198***
	(0.038)	(0.064)	(0.039)	(0.085)	(0.059)	(0.111)	(0.157)	(0.153)	(0.262)	(0.298)
Number of Dismissals * 1950	-0.246**	-0.187***	-0.343**	-0.254***	-0.416*	-0.354***	-0.719**	-0.739***	-0.740	-1.078***
	(0.09)	(0.028)	(0.157)	(0.056)	(0.221)	(0.099)	(0.321)	(0.122)	(0.467)	(0.309)
Number of Dismissals * 1961	-0.286***	-0.191***	-0.433***	-0.237***	-0.539***	-0.345***	-0.798***	-0.662***	-0.918***	-1.189***
	(0.071)	(0.039)	(0.102)	(0.061)	(0.149)	(0.094)	(0.258)	(0.148)	(0.291)	(0.280)
Number of Dismissals * 1970	-0.323***	-0.215***	-0.498***	-0.278***	-0.621***	-0.428***	***968.0-	-0.845***	-1.136***	-1.441***
	(0.069)	(0.050)	(0.088)	(0.097)	(0.130)	(0.113)	(0.265)	(0.145)	(0.295)	(0.296)
Number of Dismissals * 1980	-0.272***	-0.181***	-0.399***	-0.237	-0.546***	-0.417**	-0.730***	-0.764***	-1.019***	-1.583***
	(0.061)	(0.064)	(0.104)	(0.160)	(0.125)	(0.155)	(0.259)	(0.184)	(0.270)	(0.381)
Of Destamostics * Vess Dessession	Ç			5	0		Ç	50	5	
% Destruction Tear Dummies	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Department FE	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
${ m Subject}^*{ m Year}$ FE	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Occupation Zones * Post45	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
% City Destruction * Year Dummies	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
All additional controls	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Observations	714	714	714	714	714	714	714	714	714	714
R-squared	0.718	0.552	0.726	0.548	0.720	0.555	0.708	0.572	0.683	0.584
***significant at 1% **significant at 5%	at 5%	*significant at	at 10%	(s.e. clustered	(s.e. clustered at university level)	evel)				

The dependent variable Publications reported in odd columns is the sum of publications published by all scientists in department d in a five-year window around year t. The dependent variable Citation weighted Pubs. reported in even columns is the sum of publications published by all scientists in department d in a five-year window around year t. Dependent variables are normalized to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals in Nazi Germany between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * Year Dummies* is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with a set of year indicatores as in the main specification. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6. (s.e. clustered at university level) *significant at 10% **significant at 5%

Table 9: Persistence of Dismissal and Bombing Shocks - Department Size

	(1)	(2)	(3)
	Department	Department	Department
Dependent Variable:	Size	Size	Size
Number of Dismissals * 1926	-0.217***		-0.218***
Trumber of Dismissais 1920	(0.060)		(0.056)
Number of Dismissals * 1940	-0.612***		-0.604***
Trainibel of Bishibbalo 1010	(0.078)		(0.083)
Number of Dismissals * 1950	-1.129***		-1.124***
Training of Biblingball 1000	(0.353)		(0.336)
Number of Dismissals * 1961	-1.045***		-1.064***
	(0.267)		(0.257)
Number of Dismissals * 1970	-0.935**		-1.009**
	(0.381)		(0.374)
Number of Dismissals * 1980	0.063		0.024
	(0.665)		(0.656)
% Destruction * 1926		0.005	0.006
70 2 0501 4001011		(0.013)	(0.013)
% Destruction * 1931		0.005	0.004
,, =		(0.011)	(0.010)
% Destruction * 1950		-0.034	-0.029
		(0.025)	(0.024)
% Destruction * 1961		-0.008	-0.003
		(0.029)	(0.029)
% Destruction * 1970		0.050*	0.056**
		(0.025)	(0.027)
% Destruction * 1980		$0.034^{'}$	0.041
		(0.055)	(0.056)
Department FE	yes	yes	yes
Subject*Year FE	yes	yes	yes
Occupation Zones * Post45	yes	yes	yes
% City Destruction * Year Dummies	yes	yes	yes
All additional controls	yes	yes	yes
Observations	714	714	714
R-squared	0.882	0.878	0.884
	nificant at 5% b		ignificant at 1

^{***}significant at 1% level

(All standard errors clustered at university level)

The dependent variable Department Size measures department size in department d and year t. Number of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * 1926 is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is % destruction with 1940, the last observation before the bombings. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6.

^{**}significant at 5% level

^{*}significant at 10% level

Table 10: Quality of New Hires

	(1)	(2)	(3)	(4)	(5)
	Quality	Quality	Quality	Quality	Quality
Dependent Variable:	of Hires	of Hires	of Hires	of Hires	of Hires
	All	Above	Top 25th	Top 10th	Тор
Dismissal Quality:	Dismissals	median Quality	percentile	percentile	5th percentile
Number of Dismissals * 1940	-0.059***	-0.096***	-0.145***	-0.249***	-0.428***
Trumber of Dismissais 1940	(0.014)	(0.018)	(0.040)	(0.051)	(0.133)
Number of Dismissals * 1950	-0.055**	-0.090***	-0.138***	-0.234***	-0.400***
Trains of ST	(0.020)	(0.027)	(0.046)	(0.061)	(0.139)
Number of Dismissals * 1961	-0.054***	-0.089***	-0.134***	-0.226***	-0.417***
	(0.013)	(0.018)	(0.037)	(0.049)	(0.120)
Number of Dismissals * 1970	-0.055***	-0.090***	-0.141***	-0.266***	-0.484***
	(0.020)	(0.028)	(0.041)	(0.054)	(0.130)
Number of Dismissals * 1980	-0.064***	-0.112***	-0.171***	-0.285***	-0.520***
	(0.017)	(0.031)	(0.054)	(0.069)	(0.160)
% Destruction * Year Dummies	ves	yes	yes	yes	yes
Department FE	yes	yes	yes	yes	yes
Subject*Year FE	yes	yes	yes	yes	yes
Occupation Zone * Post45	yes	yes	yes	yes	yes
% City Destruction * Year Dummies	yes	yes	yes	yes	yes
All additional controls	yes	yes	yes	yes	yes
Observations	602	602	602	602	602
R-squared	0.401	0.406	0.421	0.436	0.453

The dependent variable Quality of Hires measures the average quality of new hires in department d between year t and year t-1. Quality of hires is measured by the career average of citation-weighted publications. The variable is normalized to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * Year Dummies is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with a full set of year dummies as in the main specification. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6.

8 Appendix

8.1 Appendix Figures

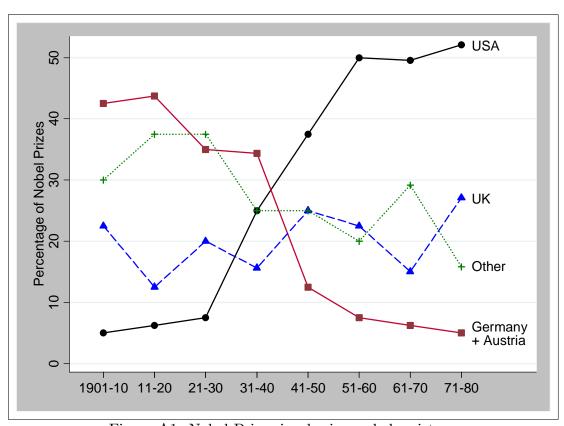


Figure A1: Nobel Prizes in physics and chemistry

Note: The figure reports the percentage of Nobel Prizes awarded in physics and chemistry to scientists affiliated with a university in the respective country for each decade from 1901 to 1980. Prizes are weighted according to the fractions awarded by the Nobel committee (i.e. if the prize was awarded to 3 scientists in one year with one scientist getting 0.5 of the prize, and the other two scientists receiving 0.25 of the prize their countries would be assigned 0.5 and 0.25 respectively). Over the time period 1901 to 1980 scientists based in Austrian universities contribute 2 prizes to the combined total of 33.75 prizes awarded to scientists in German and Austrian universities. Data on Nobel Prizes and university affiliations come from http://www.nobelprize.org/nobel_prizes/.

8.2 Appendix Tables

Table A1: Top Journals

Journal Name	Published in	Historical Top Journal	Current Top Journa
General Journals			
Nature	UK	yes	yes
Naturwissenschaften	Germany	yes	y CB
Proceedings of the National Academy of Sciences	USA	<i>y</i> 05	yes
Proceedings of the Royal Society of London A (Mathematics and Physics)	UK	yes	J CB
Science	USA	yes	yes
Sitzungsberichte der Pressischen Akademie der Wissenschaften	Germany	yes	yes
Physics			
Annalen der Physik	Germany	yes	
Applied Physics Letters	USA	J	yes
Astrophysical Journal	UK		yes
Journal of Applied Physics	USA		yes
Journal of Chemical Physics	USA		yes
Journal of Geophysical Research B: Solid Earth	USA		yes
Physical Review	USA	yes	yes
Physical Review A	USA	yes	
Physical Review B	USA		yes
			yes
Physical Review C	USA		yes
Physical Review D	USA		yes
Physical Review Letters	USA		yes
Physikalische Zeitschrift	Germany	yes	
Zeitschrift für Physik	Germany	yes	
Chemistry	TICA		
Analytical Chemistry	USA		yes
Angewandte Chemie - International Edition in English	UK		yes
Berichte der Deutschen Chemischen Gesellschaft	Germany	yes	
Biochemische Zeitschrift	Germany	yes	
Chemical Communications	USA		yes
Inorganic Chemistry	USA		yes
Journal für Praktische Chemie	Germany	yes	
Journal of Biological Chemistry	USA		yes
Journal of Organic Chemistry	USA		yes
Journal of Physical Chemistry	USA	yes	yes
Journal of the American Chemical Society	USA		yes
Journal of the Chemical Society	UK	yes	
Justus Liebigs Annalen Chemie	Germany	yes	
Kolloid Zeitschrift	Germany		
Tetrahedron Letters	Netherlands		yes
Zeitschrift für Anorganische Chemie und Allgemeine Chemie	Germany	yes	
Zeitschrift für Elektrochemie und Angewandte Physikalische Chemie	Germany	yes	
Zeitschrift für Physikalische Chemie	Germany	yes	
Mathematics			
Acta Mathematica	Sweden	yes	yes
Advances in Mathematics	USA		yes
Annals of Mathematics	USA	yes	yes
Bulletin of the American Mathematical Society	USA		yes
Inventiones Mathematicae			yes
Jounal für die reine und angewandte Mathematik	Germany	yes	=
Journal of Functional Analysis	USA	-	yes
Journal of the London Mathematical Society	Germany		J
Mathematische Annalen	Germany	yes	
Mathematische Zeitschrift	Germany	yes	
Philosophical Transactions of the Royal Society A	UK	ycs	7700
Proceedings of the London Mathematical Society	UK	WOO	yes
L LOCECOTORS OF THE LONGON IVIAL HEINALICAL SOCIETY	\cup IX	yes	

Table A2: Top Scientists

			Yearly Career		Dis-	First	Last
Ν	TT :'/ 1	II ::4 0	Cit. weighted	Nobel	missed	year	year
Name Physics	University 1	University 2	Publications	Prize	1933-40	in data	in data
Wigner, Eugen	Berlin TU		619.8	7700	1933	1931	1931
Binder, Kurt	Köln		468.3	yes	1999	1980	1980
Cardona, Manuel	Stuttgart TU		284.3			1980	1980
Ewald, Peter Paul	Stuttgart TU		161.8		1937	1926	1931
Wegner, Franz	Heidelberg		148.3		1001	1980	1980
Born, Max	Göttingen		144.2	yes	1933	1926	1931
Greiner, Walter	Frankfurt		135.6	<i>J</i> 02	1000	1970	1980
Schrödinger, Erwin	Berlin		129.6	yes	1933	1926	1931
Schmidt, Michael	Heidelberg		112.5	J		1980	1980
Bergmann, Gerd	Köln		97.3			1980	1980
Haken, Hermann	Stuttgart TU		96.5			1961	1980
Hess, Karl	Wien		91.5			1980	1980
Schmid, Albert	Karlsruhe TU		88.2			1970	1980
Hohenberg, Pierre	München TU		87.9			1980	1980
Einstein, Albert	Berlin		82.2	yes	1933	1926	1931
Schatz, Gerd	Heidelberg		73.5			1980	1980
Müller, Bernd	Frankfurt		70.1			1980	1980
Fulde, Peter	Frankfurt	Darmstadt TU	68.4			1970	1980
Schlögl, Friedrich	Aachen TU		67.2			1961	1980
Gross, Ferdinand	Graz		66.2			1970	1980
Clara intern							
Chemistry Mayorhof Otto	Uoidell on :		977 4		1020	1091	1091
Meyerhof, Otto	Heidelberg		277.4	yes	1938	1931	1931
Sies, Helmut	München		172.6 163.5		1938	1980	1980
Neuberg, Carl	Berlin Münghon			*****	1950	1926	1931
Lynen, Feodor Eckstein, Fritz	München Göttingen		160.2 159.2	yes		1961 1980	$1970 \\ 1980$
Giese, Bernd	Darmstadt TU		153.0			1980	1980
Reetz Manfred T.	Marburg		151.0			1980	1980
Pette, Dirk	München		141.1			1970	1970
Lohmann, Karl	Heidelberg	Berlin	136.1			1931	1961
Neupert, Walter	München	Dermi	135.8			1980	1980
Bergmann, Max	Dresden TU		129.6		1933	1926	1931
Vorbrüggen, Helmut	Berlin TU		125.2		1000	1980	1980
von Raque Schleyer	Erlangen		110.8			1980	1980
Paulsen, Hans	Hamburg		110.0			1970	1980
Witkop, Bernhard	München		108.9			1950	1950
Hoppe, Rudolf	Gießen		106.3			1961	1980
Vögtlke, Fritz	Würzburg		104.7			1980	1980
Kessler, Horst	Frankfurt		103.8			1980	1980
Wieghardt, Karl	Hannover TU		95.0			1980	1980
Westphal, Otto	Freiburg		94.2			1961	1980
Mathematics							
von Neumann, Johann	Berlin		150.6		1933	1931	1931
Keller, Wilfried	Hamburg		75.6			1980	1980
Bott, Raoul	Bonn		51.8			1961	1970
Kaup, Wilhelm	Tübingen		43.2			1980	1980
Lorentz, George G.	Tübingen		39.7			1950	1950
von Mises, Richard	Berlin TU		38.2		1933	1926	1931
Friedrichs, Kurt	Göttingen		37.4		1937	1931	1931
Jensen, Ronald	Bonn		35.6			1980	1980
Krieger, Wolfgang	Heidelberg		35.3			1980	1980
Barth, Wolf	Erlangen		29.1		1099	1980	1980
Szegő, Gabriel	Berlin		27.6		1933	1926	1931
Löh, Hans-Günter	Hamburg		26.2		1022	1980	1980
Weyl, Hermann	Göttingen		26.0		1933	1926	1931
Schaeffer, Helmut	Hamburg Cöttingen		24.1		1933	1980	1980
Lewy, Hans	Göttingen Heidelberg		23.4		1933	1931	1931
Dold, Albrecht Grauert, Hans	Heidelberg Göttingen		$\frac{22.3}{18.7}$			$1961 \\ 1961$	1980
Becker, Jochen	Göttingen Berlin TU		18.3			1981	1980 1980
Hausdorff, Felix	Bonn 10		16.3 16.7			1980	1980
Menger, Karl	Wien		16.7		1938	1926	1931
menger, Karr	4 A 1G11		10.7		1390	1391	1391

Table A3: First Stages - Instrumenting with University Destruction for Subject

<u>Destruction</u> (1) $\overline{(2)}$ (3)(4) % Subj. % Subj. % Subj. % Subj. % Subj. % Subj. Destruction Destruction Destruction Destruction Destruction Destruction Dependent Variable: * 1926 * 1931 * 1950 * 1961 * 1970 * 1980 0.602*** % Uni. Destruction * 1926 -0.011 0.0030.0030.0030.005(0.004)(0.011)(0.166)(0.014)(0.004)(0.006)0.602*** % Uni. Destruction * 1931 -0.011 0.0030.003 0.003 0.004(0.013)(0.167)(0.004)(0.004)(0.005)(0.008)% Uni. Destruction * 1950 0.608*** -0.001 0.001-0.002 0.0010.015(0.026)(0.026)(0.148)(0.024)(0.023)(0.017)0.611*** % Uni. Destruction * 1961 -0.000 0.001-0.0050.001 0.014(0.026)(0.026)(0.025)(0.148)(0.023)(0.020)0.611*** % Uni. Destruction * 1970 0.0010.004-0.013 -0.0080.026(0.027)(0.027)(0.027)(0.025)(0.148)(0.031)% Uni. Destruction * 1980 -0.021 0.715*** 0.0020.005-0.041-0.034(0.040)(0.039)(0.060)(0.061)(0.070)(0.250)Number of Dismissals * Year Dummies ves yes yes yes ves yes Department FE yes yes yes yes yes yes Subject*Year FE ves ves ves ves ves ves Occupation Zones * Post45 yes yes yes yes yes yes % City Destruction * Year Dummies yes yes yes yes yes ves All additional controls yes yes yes yes yes yes Observations 714714714714714714

0.814

R-squared

Cragg-Donald EV Statistic

The dependent variable % Dep. Destruction * 1926 reported in column (1) is equal to percentage destruction at the department level caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. Dependent variables in columns (2) to (6) are defined accordingly. The instrumental variable % Uni. Destruction * 1926 is equal to percentage destruction at the university level caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other instrumental variables are defined accordingly. The control variables Number of Dismissals * Year Dummies are equal to the number of dismissals in Nazi Germany between 1933 and 1940 interacted with a full set of year dummies. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6.

0.813

0.810

19.6

0.810

0.811

0.865

Table A4: Interaction of Human and Physical Capital

	(1)	(2)	(3)	(4)
	` '	1 1	Citation	Citation
			weighted	weighted
Dependent Variable:	Publications	Publications	Publications	Publications
Number of Dismissals * 1926	0.026	0.026	-0.087	-0.087
	(0.026)	(0.026)	(0.077)	(0.077)
Number of Dismissals * 1940	-0.175***	-0.174***	-0.181***	-0.181***
	(0.038)	(0.039)	(0.064)	(0.064)
Number of Dismissals * 1950	-0.246**	-0.159	-0.187***	-0.162***
	(0.099)	(0.096)	(0.028)	(0.035)
Number of Dismissals * 1961	-0.286***	-0.271***	-0.191***	-0.174***
	(0.071)	(0.091)	(0.039)	(0.041)
Number of Dismissals * 1970	-0.323***	-0.316***	-0.215***	-0.244***
	(0.069)	(0.084)	(0.050)	(0.080)
Number of Dismissals * 1980	-0.272***	-0.284***	-0.181***	-0.222**
	(0.061)	(0.072)	(0.064)	(0.106)
% Destruction * 1926	-0.002	-0.002	-0.003	-0.003
	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1931	-0.005*	-0.005*	-0.003	-0.003
	(0.002)	(0.002)	(0.004)	(0.004)
% Destruction * 1950	-0.005**	-0.002	-0.004*	-0.003
	(0.002)	(0.002)	(0.002)	(0.002)
% Destruction * 1961	-0.003	-0.002	-0.002	-0.002
	(0.002)	(0.002)	(0.003)	(0.003)
% Destruction * 1970	0.003	0.004	0.007	0.006
	(0.002)	(0.003)	(0.004)	(0.004)
% Destruction * 1980	0.001	0.000	0.009	0.007
	(0.003)	(0.004)	(0.007)	(0.006)
Number of Dismissals * % Destruction * 1950		-0.003**		-0.001
		(0.001)		(0.001)
Number of Dismissals * % Destruction * 1961		-0.000		-0.001
		(0.001)		(0.001)
Number of Dismissals * % Destruction * 1970		-0.000		0.001
		(0.001)		(0.002)
Number of Dismissals * % Destruction * 1980		0.001		0.002
		(0.002)		(0.003)
Department FE	yes	yes	yes	yes
Subject*Year FE	yes	yes	yes	yes
Occupation Zones * Post45	yes	yes	yes	yes
% City Destruction * Year Dummies	yes	yes	yes	yes
All additional controls	yes	yes	yes	yes
Observations	714	714	714	714
R-squared	0.718	0.723	0.552	0.555
**significant at 1% **significant at 5%		cant at 10%		d at university

^{***}significant at 1% **significant at 5% *significant at 10% (s.e. clustered at university level)

The dependent variable Publications is the sum of publications published by all scientists in department d in a five-year window around year t. The dependent variable Citation weighted Pubs. is the sum of publications published by all scientists in department d in a five-year window around year t. Dependent variables are normalized to have zero mean and a standard deviation of one within subjects. Number of Dismissals * 1926 is equal to the number of dismissals between 1933 and 1940 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is the number of dismissals with 1931, the last observation before the dismissals. % Destruction * 1926 is equal to percentage destruction caused by Allied bombings between 1940 and 1945 interacted with an indicator that is equal to 1 for observations from 1926. The other interactions are defined accordingly. The excluded interaction is % destruction with 1940, the last observation before the bombings. Number of Dismissals * % Destruction * 1950 is the triple interaction of the number of dismissals, percentage destruction, and an indicator for 1950. The other triple interactions are defined accordingly. Department FE is the full set of 105 department fixed effects. Subject*Year FE is the interaction of subject indicators (for physics, chemistry, and mathematics) with year dummies. Occupation Zones * Post45 is the interaction of occupation zone indicators (English zone, French zone, Soviet zone, excluded category U.S. zone) with a post-1945 dummy. % City Destruction * Year Dummies is the interaction of city level destruction with the full set of year dummies. All additional controls is the full set of additional control variables as used in column (6) of Tables 5 and 6.

8.3 Calculation of Contribution of Human and Physical Capital Shocks for the Decline of German Science

Dismissal Shock

The effect of the dismissals on German science is calculated using the regression results including all controls as in column (6) of Tables 5 and 6. Using the number of dismissals in each department I calculate the reduction in (citation weighted) publications in each department for 1940, 1950, 1961, 1970, and 1980 in terms of standard deviations. For each department and year I therefore calculate:

$$\Delta y_{1940} = \widehat{\beta}_{1940}^{\textit{dismissals}} * (\# \text{ of Dismissals 33-40}), ..., \ \Delta y_{1980} = \widehat{\beta}_{1980}^{\textit{dismissals}} * (\# \text{ of Dismissals 33-40})$$

Multiplying the Δy 's with the subject level standard deviations of (citation weighted) publications I calculate the fall in productivity in each department in terms of (citation weighted) publications (call them ΔY_{year}). The ΔY_{year} 's compute the reduction in (citation weighted) publications for the years 1940, 1950, 1961, 1970, and 1980. To obtain the total reduction in (citations weighted) publications for all years since 1933, I assume that the productivity loss between April 1933 and December 1945 was ΔY_{1940} in each year. Similarly, between January 1946 and December 1955 the annual loss in productivity was ΔY_{1950} , and so on. Total productivity loss between 1933 and 1980 was therefore:

$$\Delta Y_{1933-1980} = 12.75 * \Delta Y_{1940} + 10 * \Delta Y_{1950} + 10 * \Delta Y_{1961} + 10 * \Delta Y_{1970} + 5 * \Delta Y_{1990}$$

Adding $\Delta Y_{1933-1980}$'s for all departments in a subject, I obtain the total loss in (citation weighted) publications in each subject from 1933 to 1980 ($\Delta Y_{1933-1980}^{all}$).

To calculate percentage losses I obtain the total number of (citation weighted) publications that were published in a subject in 1940, 1950, 1961, 1970, and 1980:

$$Y_{1940}^{tot}, Y_{1950}^{tot}, ..., Y_{1980}^{tot}$$

Average yearly (citation weighted) publications are obtained as follows:

$$Y_{yearly}^{tot} = \frac{1}{5} (\ Y_{1940}^{tot} + \ Y_{1950}^{tot} + \ Y_{1961}^{tot} + \ Y_{1970}^{tot} + \ Y_{1980}^{tot})$$

Total publications between April 1933 and December 1980 are calculated as:

$$Y_{1933-1980}^{tot} = 47.75 * Y_{vearly}^{tot}$$

Finally, percentage loss between 1933 and 1980 is calculated as:³⁸

$$\%\Delta Y_{1933-1980}^{all} = \frac{\Delta Y_{1933-1980}^{all}}{(Y_{1933-1980}^{tot} - \Delta Y_{1933-1980}^{all})} * 100$$

The top panel of Table A5 summarizes the total loss of (citation weighted) publications between 1933 and 1980 that was caused by the dismissal of scientists in Nazi Germany.

Bombing Shock

I calculate the effect of Allied bombings on German science in a similar way. The calculations also rely on the regression results including all controls as in column (6) of Tables 5 and

³⁸Note: $\Delta Y_{1933-1980}^{all} < 0$.

6. Using percentage destruction in each department I calculate the reduction in (citation weighted) publications in each department for 1950, 1961, 1970, and 1980 in terms of standard deviations. For each department and year I therefore calculate:³⁹

$$\Delta y_{1950} = \widehat{\beta}_{1950}^{bombings} * (\% \text{ Destruction 42-45}),..., \ \Delta y_{1980} = \widehat{\beta}_{1980}^{bombings} * (\% \text{ Destruction 42-45})$$

Multiplying the Δy 's with the subject level standard deviations of (citation weighted) publications I calculate the fall in productivity in each department in terms of (citation weighted) publications (call them ΔY_{year}). The ΔY_{year} 's compute the reduction in (citation weighted) publications for the years 1950, 1961, 1970, and 1980. To obtain the total reduction in (citations weighted) publications for all years since 1944, I assume that the productivity loss between January 1944 and December 1955 was ΔY_{1950} in each year. Similarly, between January 1956 and December 1965 the annual loss in productivity was ΔY_{1961} , and so on. Total productivity loss between 1944 and 1980 is therefore:

$$\Delta Y_{1944-1980} = 11 * \Delta Y_{1950} + 10 * \Delta Y_{1961} + 10 * \Delta Y_{1970} + 5 * \Delta Y_{1990}$$

Adding the $\Delta Y_{1944-1980}$'s for all departments in a subject, I obtain the total loss in (citation weighted) publications that was caused by Allied bombings in each subject from 1944 to 1980 $(\Delta Y_{1944-1980}^{all})$.

To calculate percentage losses I obtain the total number of (citation weighted) publications that were published in a subject in 1950, 1961, 1970, and 1980:

$$Y_{1950}^{tot},\,Y_{1961}^{tot},\!...,\,Y_{1980}^{tot}$$

Average yearly (citation weighted) publications are obtained as follows:

$$Y_{yearly}^{tot} = \frac{1}{4}(Y_{1950}^{tot} + Y_{1961}^{tot} + Y_{1970}^{tot} + Y_{1980}^{tot})$$

Total publications between January 1944 and December 1980 are calculated as:

$$Y_{1944-1980}^{tot} = 36 * Y_{yearly}^{tot}$$

Finally, percentage loss between 1944 and 1980 is calculated as:⁴⁰

$$\%\Delta Y_{1944-1980}^{all} = \frac{\Delta Y_{1944-1980}^{all}}{(Y_{1944-1980}^{tot} - \Delta Y_{1944-1980}^{all})} * 100$$

The bottom panel of Table A5 summarizes the total loss of (citation weighted) publications between 1944 and 1980 that was caused by Allied bombings.

 $[\]overline{}^{39}$ I only consider $\widehat{\beta}_x^{bombings}$ if the coefficient is at least significant at the 10 percent level for year X. For all other years I set $\Delta y = 0$. As a result, $\Delta y_{1961}, \Delta y_{1970}, \Delta y_{1980}$ are set to 0. See column (6) of Tables 5 and 6. $\overline{}^{40}$ Note: $\Delta Y_{1933-1980}^{all} < 0$.

Table A5: Total Productivity Loss of Dismissals and Bombings

•			G	
	Physics	Chemistry	Mathematics	Total
Dismissal Loss				
Number of publications lost 1933-1980	2029	6848	699	9576
Number of citation weighted publication lost 1933-1980	60703	122248	8969	191920
Percentage of publications lost 1933-1980	30.5	36.5	33.5	33.5
Percentage of citation weighted publications lost 1933-1980	34.0	33.2	36.6	34.6
Publications by dismissed scientists	362	594	225	1181
Citation weighted publications by dismissed	14826	12708	4835	32369
Bombing Loss				
Number of publications lost 1944-1980	231	710	87	1028
Number of citation weighted publication lost 1944-1980	7410	13589	1195	22194
Percentage of publications lost 1944-1980	5.0	5.9	6.2	5.7
Percentage of citation weighted publications lost 1944-1980	6.2	5.5	7.6	6.4

8.4 Data Appendix

8.4.1 Panel Data Set of Scientists in German and Austrian Universities from 1926 to 1980

As described in the main text I use "Kürschners Deutscher Gelehrtenkalender" (KDG) to construct a panel data set of German scientists at 7 points in time between 1926 and 1980. The KDG covers all researchers in German speaking universities. To compile the KDG the editors contacted all German speaking universities to obtain faculty rosters and then sent out questionnaires to all faculty members. The response rate to these questionnaires was very high. If a scholar did not answer the questionnaire the editors of the KDG tried to find as much information as possible on the scholar.

There sometimes was a slight delay until a young researcher was included in the KDG or until a university change was recorded. A privatdozent, for example, may have been appointed in 1926 but may not appear in the 1926 volume because she was not a privatdozent at the time the questionnaires were sent out. The same scientist, however, would appear in the 1931 volume with her complete appointment history. If that history indicates that she had already been a privatdozent in 1926 I also include her in the 1926 roster. This gives a more accurate picture of each department's faculty in the relevant year.

The KDGs list researchers who occupied different university positions. I focus on all researchers who had the right to teach ('venia legendi') at a German university, i.e. all researchers who were at least privatdozent. The data therefore include ordinary professors, extraordinary professors, honorary professors, and 'privatdozenten'. The Nazi government renamed the 'privatdozent' position into 'dozent' which affects the data in 1941. To have a comparable set of researchers across different years I also add all dozenten to the data.

8.4.2 Productivity Measures for German and Austrian Science Departments

The publications and citations data cover historical and current top science journals and were downloaded from the ISI Web of Science. The set of journals is based on historical accounts of relevant top journals and on current journal rankings.

Historical top journals

The list of top journals in the 1920s and 1930s includes mostly German journals but also the major international journals. As German science was leading at the time, many of the German journals were among the best journals worldwide which is underlined by an article published in Science in 1941: "Before the advent of the Nazis the German physics journals (Zeitschrift für Physik, Annalen der Physik, Physikalische Zeitschrift) had always served as the central organs of world science in this domain [...] In 1930 approximately 700 scientific papers were printed in its [the Zeitschrift für Physik's] seven volumes of which 280 were by foreign scientists" (American Association for the Advancement of Science, 1941).

I obtain the list of historical journals in three step process. First, I obtain all German science journals published in the 1920s to 1940s that are included in the Web of Science. Second, I include three general science journals that were relevant outlets for German scientists publishing in the 1920s and 1930s: Nature, Science, and the Proceedings of the Royal Society. Finally, the list of historical top journals is augmented by four international field journals that have been recommended by historians of science as important outlets for German scientists. Relevant chemistry journals were suggested by Ute Deichmann and John Andraos who work on chemistry in the early 20th century. Historical mathematics journals were suggested by Reinhard Siegmund-Schultze and David Wilkins who are specialists in the history of mathematics.

Current top journals

The definition of top journals for German (and international) scientists changed substantially since the 1920s and 30s. To reflect this change in my productivity measure I also compile a second list of top journals based on current international journal rankings. I use rankings provided by SCImago Journal & Country Rank to obtain the ten most cited journals in general science, physics, and chemistry. SCImago does not rank mathematics journals.⁴¹ I therefore obtain the current most cited mathematics journals from a commonly used ranking provided by the University of Texas.⁴²

Universe of Articles in Top Science Journals Published Between 1920 and 1985

The overall list of top science journals includes 51 journals. I download all articles published in these journals between 1920 and 1985. I.e. even if a journal only became a top journal in later years I download all articles published in the journal since 1920. A small number of journals were only founded after 1920. For these journals I download all articles since the creation of the journal. The publication of a few journals was interrupted towards the end of WWII. As

⁴¹See http://www.scimagojr.com, accessed 13th of May 2010.

⁴²See http://www.ma.utexas.edu/users/lsilvest/rankings/mranking.html, accessed 13th of May, 2010.

a result these journals have missing data during those years. Furthermore, some journals have missing data in the Web of Science for some years even though the journal was published in that year.⁴³ The inclusion of year fixed effects in all regressions addresses this problem.

8.4.3 Data on Bombings of German Universities and Science Departments

Data on university level bombing destruction come from university websites and from Tietze (1995), Phillips (1983), Samuel and Thomas (1949), Schneider (1990), and Cheval (1991).

As outlined in the main text, data on department level bombing destruction is obtained by contacting university archivists and asking them to provide destruction information for buildings used by physicists, chemists, and mathematicians.⁴⁴ Detailed data sources for department level destruction are listed in Table A6.

If a department occupied more than one building (e.g. one building for the institute of experimental physics and a different building accommodating the institute for theoretical physics) I average percentage destruction across all buildings.

In some cases the historical sources only provide verbal descriptions of bombing destruction. I convert these descriptions into percentage destruction according to the following rule:

Verbal description	Percentage destruction
"completely destroyed"	100%
"heaviest destruction" or "destroyed to a large extent"	75%
"heavy destruction"	50%
"part destruction" or "burnt out"	25%
"light destruction"	10%

8.4.4 Data Sources of Control Variables

I obtain data on control variables from a number of sources.

Number of Departments Within 50km

For each university I calculate the number of departments in the same subject within 50km. The measure also includes universities that were founded after 1945. The full list of universities in Germany as of 2010 was obtained from "Personal and Hochschulen - Fachserie 11, Reihe 4.4, 2010" accessed online at http://www.destatis.de/jetspeed/portal/cms/Sites/destatis/Internet/

⁴³I have highlighted this problem to Thomson Scientific. It is caused by error prone scanning of historical journals.

⁴⁴The following university archivists put in a lot of time and effort to gather information on department level bombing destruction or to provide access to the relevant sources: Klaus Graf (Aachen TU), Claudia Schülzky (Berlin TU), (Berlin), Thomas Becker (Bonn), Klaus Oberdieck (Braunschweig TU), Matthias Lienert (Dresden TU), Michael Maaser (Frankfurt), Dieter Speck (Freiburg), Eva-Maria Felschow (Gießen), Ulrich Hunger (Göttingen), Alois Kernbauer (Graz), Marieluise Vesulak (Graz TU), Ralf-Torsten Speler (Halle), Eckart Krause (Hamburg), Lars Nebelung (Hannover TU), Peter Goller (Innsbruck), Joachim Bauer (Jena), Klaus Nippert (Karlsruhe TU), Dagmar Bickelmann (Kiel), Andreas Freitäger (Köln), Jens Blecher and Roy Lämmel (Leipzig), Katharina Schaal (Marburg), Hans-Michael Körner (München), Margot Fuchs (München TU), Sabine Happ (Münster), Norbert Becker (Stuttgart), Thomas Maisel (Wien), Juliane Mikoletzky (Wien TU), Marcus Holtz (Würzburg).

DE/Content/Publikationen/Fachveroeffentlichungen/BildungForschungKultur/Hochschulen/P ersonalHochschulen2110440107004,property=file.pdf, a publication of the German statistical agency (Statistisches Bundesamt). A list of Austrian universities in 2011 was obtained from the Austrian statistical agency (Statistik Austria) accessed online at: http://www.statistik.at/web_de/statistiken/bildung_und_kultur/formales_bildungswesen/universitaeten_studium/index. html. Using university websites I check for their founding year and whether they have a physics, chemistry, or mathematics department. This allows me to calculate the number of departments within 50km for each department and year in my sample.

Armament Related Industries in 1933

Data on the share of firms in three armament related industries (iron and steel production, mechanical engineering and vehicle construction, chemical industry) come from the establishment census of 1933 that was published in "Statistik des Deutschen Reichs – Band 463: Gewerbliche Betriebszählung, 1935". Data on industry shares in Austria come from the establishment census of 1930 that was published by the Bundesamt für Statistik in "Gewerbliche Betriebszählung in der Republik Österreich, 1932". The data measure the share of firms that belong to a certain industry (among all firms) at the city level.

Fraction of Jews in 1933

The fraction of Jews in 1933 is based on German census data from 1933. The data were obtained from "Statistik des Deutschen Reiches: Die Bevölkerung des Deutschen Reichs nach den Ergebnissen der Volkszählung 1933, Band 451, Heft 3 (1936)". As the German census of 1933 did not cover Austrian cities data on the Jewish population in the three Austrians cities in my sample were obtained from a number of different sources. Data for Vienna are for the year 1934 and come from "Statistisches Jahrbuch der Stadt Wien 1930-1935 Neue Folge, 3. Band". Data for Graz are from 1938 and come from "Israelitische Kultusgemeinde für Steiermark, Kärnten und die politischen Bezirke des Burgenlandes Oberwart, Güssing und Jennersdorf" and were accessed online at http://www.ikg-graz.at/. Data on Innsbruck are from 1938 and come from Salinger (2007).

Distance to the Iron Curtain

Distance to the Iron Curtain for German cities come from Redding and Sturm (2008). Distance to the Iron Curtain for Austrian cities is measured equivalently using the original Redding and Sturm method.⁴⁵

⁴⁵I thank Daniel Sturm for kindly offering to use his material to measure distances to the iron curtain for Austrian cities.

Table A6: Detailed Data Sources for Department Level Destruction Data

University	Source for Department Level Destruction
Aachen TU	Kriegsschäden Akten 438, 1189, 1234a
Berlin	Humboldt Universität Berlin, Universitätsarchiv, Bestand des Universitätskurators, Aktennr. 655
Berlin TU	Universitätsarchiv der Technischen Universität Berlin in der Universitätsbibliothek, 602-44
Bonn	van Rev (1995)
Braunschweig TU	Kuhlenkamp (1976)
Darmstadt TU	missing
Dresden TU	Technische Universität Dresden (1996)
Erlangen	no bombing destruction
Frankfurt	Universitätsarchiv Frankfurt (1947), Abteilung 50, Nr. 3046 BII, 241-244
Freiburg	Rösiger (1957)
Gießen	Universitätsarchiv Gießen, PrA. Nr. 2208
Göttingen	Brinkmann (1985)
Graz	e-mail communication with university archivist Prof. Dr. Alois Kernbauer
Graz TU	Weingand (1995), p. 58, p. 103
Greifswald	no bombing destruction
Halle	Eberle (2002)
Hamburg	e-mail communication with university archivist Eckart Krause, Kröplin (1951), pp.422-428,
	Senat Hamburg (1955), Giles (1985), p. 297
Hannover TU	Wolters (1950)
Heidelberg	no bombing destruction
Innsbruck	Klebelsberg (1953), pp. 193-196
Jena	Schmidt, Elm, Steiger, Böhlaus (1983), pp. 301-302
Karlsruhe TU	Hoepke (2007)
Kiel	Jaeger (1965), pp. 117-202
Köln	Universitätsarchiv documents
Leipzig	Füssler (1961)
Marburg	Fritzsche, Hardt, and Schade (2003), p. 30
München	Mager (1958), p. 255
München TU	Technische Hochschule München (1968)
Münster	Niemer (2010)
Rostock	no bombing destruction
Stuttgart TU	Technische Hochschule Stuttgart (1947)
Tübingen	no bombing destruction
Wien	Adamovich (1947)
Wien TU	e-mail communication with university archivist Dr. Juliane Mikoletzky
Würzburg	e-mail communication with university archivist Dr. Marcus Holtz
The table shows deta	ailed data sources for department level destruction. Detailed citations can be found below.

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