Elevating Trust by increasing Instruction hours-German G8

Reform *

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Abstract

For this project I examine the the effect of a major German high school implemented between 2001 and 2008 in most of the federal States on student's trust attitude. The reform reduced the secondary education by one year from 9 to 8 which lead to the increase in the weekly instruction hours. Since the policy G8 reform was implemented in diffrent time over the period of 7 years I use difference in difference approach with time as year of high school entry and State as fixed effects. I get a significant reult on students' trust attitude. I also tried to look into some of the potential mechanisms in line with other existing literatures.

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1 Introduction

A growing body of emperical literature has emphasized trust plays important role in economic development and direct efect on total factor productivity. (knack1997does). At the individual level, there is evidence linking trust and subjective well-being. (helliwell2010trust). Education plays a crucial role in shaping children's trust behavior in a positive way, (glaeser2007does). Raising instruction time in high school is often considered with having several positive effects for students' outcomes, especially regarding improvements in cognitive skills and performance. (huebener2017 and eventually making schooling more efficient. Even though many countries are attracted by the idea of reaping the benefits of increased instruction time in school (oecd2016), outcomes other than student performance might be affected as well. (dahmann2014impact; dahmann2018cross; hofmann2018learning).

Here aim is to analyse whether higher instruction hours affected students' self rated trust, measured as trust in people and strangers. The remainder of the paper is structured as follows. In Section2 we provide a broad literature overview on different outcomes affected by schooling and also review existing literature on the German G8-reform whereas Section3 briefly summarizes the main characteristics and features of the very reform. Afterwards, Section4 introduces the data set we use and Section5 gives reasoning on our empirical strategy.

2 Related literature

On the determinants of trust, (alesina 2002 trusts) argue that individual experiences like traumatic events or (historically rooted) discrimination as well as one's community environment are among the strongest factors reducing trust. Accordingly (glaeser 1999 social) find less trustworthy behavior between people from different races or nationalities what has also been shown by (Borgonovi 2012) in a cross-European country comparison. (dohmen 2008 representative) using 2003 and 2005 waves of the SOEP data set, that female people, elderly as well as tall ones are usually associated with exhibiting more trust. On the other hand, trust seems to be strongly connected to personality traits as shown by (dohmen 2008 representative) and (becker 2012 relationship). The former authors examine the effect of psychological traits as measured by the "Big Five" concept and argue that more conscientious or more neurotic people

trust less whereas individuals who are more agreeable or more open to experiences tend to trust more.

Moreover, there exists some evidence that schooling might affect social preferences including trust as well as altruism and reciprocity. Focusing on the effect of schooling on an individual's trust formation, current literature usually indicates that schooling and education in general have a positive impact on individual trust attitudes (see for example (oreopoulos 2011 priceless)).

3 The G8 reform

The G8 reform analyzed in this study affects only one of these tracks, the academic high school (*Gymnasium*), which constitutes the high-ability school track that upon competition leads to the Abitur, the university entrance qualification that is required for admission to the university. Typically, academic high school lasted nine years, implying a total of thirteen years of schooling. Between 2001 and 2008, 13 out of 16 German federal states reduced the length of academic track schooling from nine to eight years reducing overall schooling from 13 to 12 years. most of the additional workload usually being concentrated between grades seven to nine thus students are especially exposed to a higher workload between ages 13 and 16 (see for example (dahmann2014impact; dahmann2018cross)).

4 Data

Analysis is based on a sample of same-aged students taken from the German Socio-Economic Panel (SOEP) study, a representative household panel survey. Adolescents, who respond to the SOEP youth questionnaire in the year they turn seventeen, answer survey questions relating to trust preferences in every wave starting in 2006. Hence, restricted to the data from 2006 through 2018, and select all adolescents who were attending academic high school (Gymnasium) at the time of the survey or had earned a high school diploma.

To identify whether a student is subject to higher school intensity, used the information on the federal state of residence and the year of high-school entry. In case information on the latter is

¹Use data from the SOEP, data for years 1984-2020, version 37, SOEP, 2021.

 $^{^2}$ Exclude data of recent immigrants and refugee samples. Also exclude data from 2019 and 2020 since some states swished back to the G9 scheme

not provided, the year of high-school entry is imputed from the date of birth. Exclude students from Rhineland-Palatine where the reform has not been implemented state-wide, and students from Hesse who entered academic high school in 2004 and 2005 when schools operated under both schemes (G8 and G9). Individuals who repeated one or more grades is also excluded to avoid noise from different levels of schooling experienced.

Trust. Standard trust questions were included using three statements about whether "people can be generally trusted," whether "nowadays one can't rely on anyone," and whether "if dealing with strangers, it is better to be careful before trusting them." Answers were given on a seven-point Likert scale ranging from "agree completely" to "totally disagree." collapsed multiple measures of trust into one index and standardized it so that, by construction, trust has a mean zero and a standard deviation of one.

5 Empirical strategy

In order to estimate a causal effect of schooling intensity on trust, we exploit the aforementioned G8 reform, in particular, we exploit the fact that the reform was implemented at different points in time across federal states. We apply a difference-in-differences strategy of the following form

$$y_{ist} = \gamma_s + \tau_t + \alpha G 8_{st} + \beta X_{ist} + \epsilon_{ist}$$

where y_{ist} is the degree of trust at age 17 of student *i* living in state *s* who has entered high school at time *t*. $G8_{st}$ is a binary variable that identifies whether the student was affected by the G8 reform. Control group are students who entered academic high school before the reform yeras and graduate after nine years of high school, i.e., facing a lower schooling intensity. In contrast, our treatment group comprises students who entered academic high school after the reform and graduate after only eight years of academic high school, i.e., facing a higher schooling intensity. α is the coefficient of core interest and provides the reform's effect on student's trust. γ_s is a set of state fixed effects that captures general differences between states, like time constant differences in state's education systems. A set of time fixed effects

 (τ_t) capture general differences between cohorts over time as well as student trust shocks common to all federal states, e.g. resulting from policy changes applying to all federal states.

The set of individual control variables X_{ist} contains gender, childhood environment, previous educational performance, parental education dummy (at least one parent has an academic high school degree or higher), father's occupational status dummy (blue-collar occupation), the employment status of the mother and a dummy for being raised by a single parent, religion dummy. Finally ϵ_{ist} is the error term. As the error term is likely to be correlated within states, following the recommendation of (bertrand2004much) and cluster the standard errors at the level of the policy change.

6 Results

6.1 Graphical evidence

Figure 1 in the appendix depicts a graphical illustration of the reform's effect on trust. As the reform was implemented at different points of time, the estimated event study-style specifications with saturated leads and lags. I have used four lags periods and seven leads. Although the graph doesnot totally provide the validity of event study, I sticked to this for the purpose of their project. What can be noticed is the positive effect of trust post reform which also matches the result of main regression table.

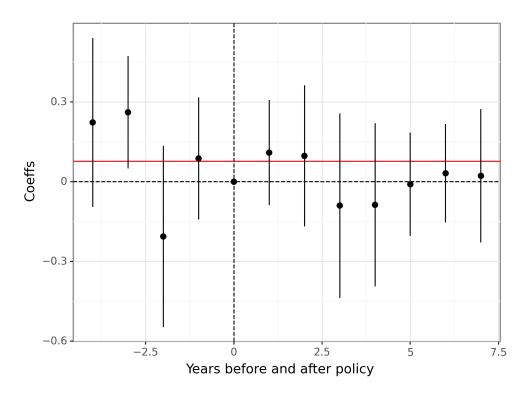


Figure 1: Event Study plot with leads and lags

6.2 Main results

Table 1 presents the estimates of equation (1). Using the controls as shown in the table I get highly significant result. The controls is similar to the paper by (dohmen 2008 representative). Although the study was not related to G8 reform Yet it showed women have more trust as contrary to the result present here. The negative effect for females on the other side could be because of the differential impact the reform had for boys and girls, with the latter often experiencing a decline in school performance (dahmann 2017 does). Significant effects for students with higher educated parents and those with a working-class father. While the former in general exhibit a substantially higher degree of trust, the latter trust less what could be related to less support these children receive at home to cope with the increased pressure at school as proposed by dahmann 2014 impactempty citation. Overall, it can be seen that the reform had an important and significantly sizeable effect on student's self rated trust.

6.3 Mechanisms

In this project I tries to look into the mechanisms which might cause the increse in trust. Tables 3,2 shows the results. Although no significant results were found may be beacuse of the lower sample size. Although postive it shows low impact for treated. There are researches which have shown low impact of the leisure time activities of children due incresed pressure. There are many other mechanisms which can be explored like the "Bive-five" as a mechanism design which I am yet to explore. Or any other in-school programes but the data set used does not contain such information.

6.4 Robust Check

Since trust was measured in the likert scale, here I used the ordered logit model to check the robustness of the main regression model, it can be expected to receive better or similar result. Table 4 shows the trust is still significant which hints to the fact the main regression is robust to model slection. There is possibility to also check the palcebo by looking at the trust attitude of the non-gymnasium students which is yet to be explored.

7 Conclusion

Drawing on data from the SOEP and using a difference-in-differences design, it is hown that the reform led to a significant increase in self-rated trust of almost 0.15 of a standard deviation with significant differential effects emerging for students with less educated parents or those being considered as low-performing. Investigate some potential mechanisms, including changes in time allocation and school-related changes and it appears to be the case that our result was driven by school-related characteristics which could not be entirely captured with our data. In addition extending gratitude to (GaudeckerEconProjectTemplates) for the project template making the work easier.

8 Appendix

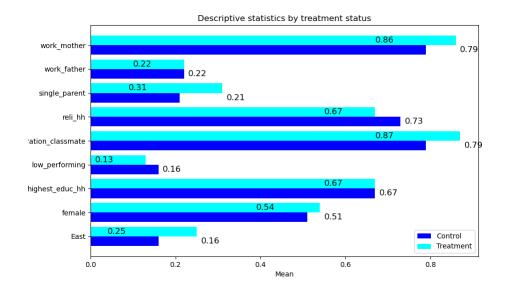


Figure 2: Descriptive statistics by treatment status

Dep. Variable: Model:	std_trust_var OLS		R-squared: Adj. R-squared:			0.075 0.053	_
Method:	Least Squares		F-statistic:			2.376e + 11	
Date:	Mon, 27 Mar 2023			Prob (F-statistic):			
Time:	17:05:20			` ikelihoo		3.53e-77 -1502.9	
No. Observations:		100	AIC:			3060.	
Df Residuals:	10	073	BIC:			3195.	
Df Model:	6	26					
Covariance Type:	clu	ıster					
		coef	std err	Z	P> z	[0.025]	0.975]
Intercept		0.4718	0.187	2.524	0.012	0.105	0.838
C(year hgsch entry)[T.	1999.0]	-0.7764	0.232	-3.353	0.001	-1.230	-0.323
C(year hgsch entry)[T.	2000.0]	-0.5513	0.223	-2.472	0.013	-0.989	-0.114
$C(year_hgsch_entry)[T.$	2001.0]	-0.4625	0.174	-2.654	0.008	-0.804	-0.121
C(year hgsch entry)[T.	2002.0]	-0.5929	0.217	-2.729	0.006	-1.019	-0.167
$C(year_hgsch_entry)[T.$	2003.0]	-0.6572	0.162	-4.061	0.000	-0.974	-0.340
$C(year_hgsch_entry)[T.$		-0.6090	0.227	-2.688	0.007	-1.053	-0.165
$C(year_hgsch_entry)[T.$		-0.6797	0.289	-2.350	0.019	-1.247	-0.113
$C(year_hgsch_entry)[T.$	_	-0.4982	0.169	-2.953	0.003	-0.829	-0.168
$C(year_hgsch_entry)[T.$		-0.7969	0.133	-6.004	0.000	-1.057	-0.537
$C(year_hgsch_entry)[T.$	_	-0.3486	0.169	-2.066	0.039	-0.679	-0.018
$C(year_hgsch_entry)[T.$		-0.5777	0.216	-2.672	0.008	-1.002	-0.154
$C(year_hgsch_entry)[T.$		-0.6464	0.203	-3.177	0.001	-1.045	-0.248
$C(year_hgsch_entry)[T.$	_	-0.5376	0.137	-3.931	0.000	-0.806	-0.270
$C(year_hgsch_entry)[T.$	_	-0.4233	0.201	-2.108	0.035	-0.817	-0.030
$C(year_hgsch_entry)[T.$	-	0.4497	0.215	2.091	0.037	0.028	0.871
$C(year_hgsch_entry)[T.$	2014.0]	0.5540	0.183	3.027	0.002	0.195	0.913
Treat		0.1537	0.054	2.851	0.004	0.048	0.259
female		-0.0556	0.056	-0.987	0.324	-0.166	0.055
East		-0.2482	0.042	-5.867	0.000	-0.331	-0.165
low_performing		0.0242	0.106	0.229	0.819	-0.183	0.231
highest_educ_hh		0.2298	0.042	5.463	0.000	0.147	0.312
work_father		-0.2871	0.070	-4.096	0.000	-0.424	-0.150
work_mother		0.1418	0.070	2.013	0.044	0.004	0.280
reli_hh		0.0594	0.057	1.044	0.297	-0.052	0.171
single_parent migration classmate		-0.0282 -0.1611	$0.080 \\ 0.093$	-0.351 -1.728	$0.726 \\ 0.084$	-0.186 -0.344	$0.129 \\ 0.022$
	4.0						U.UZZ
Omnibus:	13.948 Durbin-Watson:				1.88		
· ·	,		que-Bera	a (JB):	14.3 0.000'		
			, ,				
Kurtosis:	2.9	004 Co	nd. No.		209		

Notes:

Table 1: Python: Estimation results of the main two way fixed effect regression.

^[1] Standard Errors are robust to cluster correlation (cluster)

Dep. Variable:	volunteer_ work		R-squ	ared:		0.034		
Model:	$\overline{\mathrm{OLS}}$		Adj. I	R-square	ed:	0.025		
Method:	Least Squares		F-stat	istic:		7.335		
Date:	Mon, 27 Mar 2023		\mathbf{Prob}	(F-statis	tic): 0.	tic): 0.000491		
Time:	17:05:18		Log-L	ikelihoo	l: -698.72			
No. Observations:	1089		AIC:		1419.			
Df Residuals:	1078		BIC:			1474.		
Df Model:	10							
Covariance Type:	clus	ter						
	coef	std err	\mathbf{z}	$\mathbf{P}> \mathbf{z} $	[0.025]	0.975]		
Intercept	0.2484	0.087	2.843	0.004	0.077	0.420		
Treat	0.0129	0.019	0.668	0.504	-0.025	0.051		
female	-0.0304	0.034	-0.883	0.377	-0.098	0.037		
East	-0.0529	0.049	-1.083	0.279	-0.149	0.043		
low_performing	-0.0861	0.037	-2.326	0.020	-0.159	-0.014		
highest_educ_hh	0.0724	0.037	1.941	0.052	-0.001	0.146		
${ m work_father}$	0.0546	0.035	1.562	0.118	-0.014	0.123		
${f work_mother}$	0.0487	0.033	1.482	0.138	-0.016	0.113		
reli_hh	0.0830	0.031	2.663	0.008	0.022	0.144		
single _parent	-0.0940	0.032	-2.893	0.004	-0.158	-0.030		
$migration_classmate$	-0.0321	0.029	-1.089	0.276	-0.090	0.026		
Omnibus:	9317.043 Dur		bin-Wa	tson:	1.866			
Prob(Omnibus	s): 0.000 Jar o		que-Ber	a (JB):	174.047			
\mathbf{Skew} :	0.710 Pro l		b(JB):		1.61e-38			
Kurtosis:	1.652 Cond. No.			11.5				

Notes:

Table 2: Python: Mechanism- Volunteer work (outside school)

^[1] Standard Errors are robust to cluster correlation (cluster)

Dep. Variable:	sport_active		R-squ	ared:	(0.027		
Model:	OLS		Adj. I	R-square	ed: 0.018			
Method:	Least Squares		F-stat	istic:	34.46			
Date:	Mon,27Mar2023		Prob (F-statis		stic): 3.97e-08			
Time:	17:05:16		$\mathbf{Log}\text{-}\mathbf{L}$	ikelihoo	d: -441.99			
No. Observations:	1089		AIC:		906.0			
Df Residuals:	1078		BIC:		!	960.9		
Df Model:	10							
Covariance Type:	cluster							
	coef	std err	\mathbf{z}	P> z	[0.025]	0.975]		
Intercept	0.8846	0.049	17.937	0.000	0.788	0.981		
Treat	0.0123	0.019	0.649	0.516	-0.025	0.049		
female	-0.0532	0.017	-3.091	0.002	-0.087	-0.019		
East	-0.0942	0.038	-2.446	0.014	-0.170	-0.019		
low performing	-0.0596	0.024	-2.447	0.014	-0.107	-0.012		
highest educ hh	0.0540	0.021	2.510	0.012	0.012	0.096		
work father	-0.0243	0.035	-0.693	0.488	-0.093	0.044		
$work_mother$	0.0103	0.024	0.429	0.668	-0.037	0.057		
reli_hh	-0.0324	0.021	-1.564	0.118	-0.073	0.008		
single_parent	-0.0002	0.027	-0.009	0.993	-0.053	0.053		
migration_classmate	-0.0176	0.022	-0.796	0.426	-0.061	0.026		
Omnibus:	317.76	3 Durb	oin-Wats	son:	1.974			
Prob(Omnibus)): 0.000	Jarq	ue-Bera	(JB):	650.155			
\mathbf{Skew} :	-1.771 Prob(JB) :				6.62 e-142			
Kurtosis:	4.337 Cond. No.			11.5				

Notes:

Table 3: Python: Mechanism- School sport(inside school)

^[1] Standard Errors are robust to cluster correlation (cluster)

Dep. Variable:	${ m trust_var}$		\mathbf{Log}	Log-Likelihood:			
Model:	$\overline{\mathrm{OrderedModel}}$		AIC	AIC:			
Method:	Maximum Likelihood		od BIC	BIC:		5835.	
Date:	Mon, 27	7 Mar 2023	}				
${f Time:}$	17	:05:17					
No. Observations:	1	1100					
Df Residuals:	1	1073					
Df Model:		27					
	coef	std err	${f z}$	P> z	[0.025]	0.975]	
Treat	0.3546	0.118	3.001	0.003	0.123	0.586	
${f female}$	-0.1008	0.106	-0.952	0.341	-0.308	0.107	
East	-0.3636	0.145	-2.511	0.012	-0.647	-0.080	
$low_performing$	0.0236	0.150	0.157	0.875	-0.271	0.318	
${f highest_educ_hh}$	0.4360	0.123	3.538	0.000	0.194	0.677	
${f work_father}$	-0.5144	0.140	-3.667	0.000	-0.789	-0.239	
${f work_mother}$	0.1865	0.143	1.305	0.192	-0.094	0.467	
reli_hh	0.0557	0.122	0.457	0.648	-0.183	0.295	
${f single_parent}$	-0.0341	0.119	-0.287	0.774	-0.267	0.199	
${\it migration_class}{\it mate}$	-0.2790	0.156	-1.783	0.075	-0.586	0.028	
3/4	-4.8974	0.446	-10.977	0.000	-5.772	-4.023	
4/5	-0.1075	0.325	-0.330	0.741	-0.745	0.530	
${f 5/6}$	-0.6704	0.303	-2.213	0.027	-1.264	-0.077	
6/7	-0.5363	0.219	-2.452	0.014	-0.965	-0.108	
7/8	-0.6662	0.182	-3.670	0.000	-1.022	-0.310	
8/9	-0.6282	0.142	-4.415	0.000	-0.907	-0.349	
$\mathbf{9/10}$	-0.5582	0.111	-5.032	0.000	-0.776	-0.341	
$\boldsymbol{10/11}$	-0.7991	0.107	-7.477	0.000	-1.009	-0.590	
${\bf 11/12}$	-0.8097	0.097	-8.385	0.000	-0.999	-0.620	
12/13	-0.6514	0.083	-7.890	0.000	-0.813	-0.490	
13/14	-0.6718	0.082	-8.178	0.000	-0.833	-0.511	
14/15	-0.6276	0.084	-7.476	0.000	-0.792	-0.463	
15/16	-0.5318	0.090	-5.928	0.000	-0.708	-0.356	
16/17	-0.5274	0.108	-4.876	0.000	-0.739	-0.315	
17/18	-0.1924	0.121	-1.585	0.113	-0.430	0.045	
18/19	-0.0462	0.170	-0.272	0.786	-0.379	0.287	
$\frac{19/20}{}$	-0.1163	0.274	-0.424	0.671	-0.654	0.421	

Table 4: Python: Robustcheck