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Turning 18: What a Difference Application of Adult Criminal Law Makes

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Abstract: This paper contributes to the literature on specific deterrence by addressing the issue of selecting adolescents into adult and juvenile law systems. In Germany, different from the U.S. and most other countries, turning a critical cutoff age does not cause a sharp discontinuity from juvenile to adult penal law, but rather implies a shift to a discretionary system of both adult and juvenile law, dependent on the courts' impression of moral and mental personal development of the adolescent at the time of the act. The German legal system draws the line of adulthood at some fuzzy age interval between 18 and 21, which is well above the thresholds prevailing in the U.S. (16 to 18 years, state specific) and other countries such that the German evidence entails some external evidence to the previous literature mostly relying on U.S. data. Based on a unique inmate survey and two-equation models controlling for selectivity problems, results show that application of adult criminal law instead of juvenile penal law decreases expected recidivism of adolescents.

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1 Introduction

There is an ongoing debate on the appropriate way of punishing adolescent offenders. In the U.S., in response to increasing violence of minors during the 1990s, nearly all states moved to ease the process of transferring serious juvenile offenders to the adult criminal justice system (Myers 2003, Bishop 2000). The criminal policy of other nations is more reluctant in applying the harsher adult criminal law to young criminals. Based on the literature on personal development, opponents of the “get tough” movement argue that it seems to be more beneficial to put them back on the "right track" by using educational and disciplinary measures, as youths are generally more inclined towards criminal activity (e. g. Heinz 2004, Thornberry, Huizinga, and Loeber 2004), which can be explained by a contemporary maturity gap (Moffitt 1993). This leads to the belief that juveniles are more rehabilitable and less culpable than adults (Mears, Hay, Gertz, and Mancini 2007). However, recent developments showing an increasing number of highly aggressive and serious acts of violence committed by minors question the ruling criminal policy (see Aebi 2004, Oberwittler and Hoefler 2005). In Germany, the ongoing media coverage of violent crimes resulted in a strong pressure on politics (Bundestag 2009) to touch upon the question on how to deal with juvenile and adolescent offenders.

A major drawback of the existing empirical studies is the strong limitation to U.S. data, because U.S. courts treat individuals as adults already at an early age. 47 states use 17 or 18 as the age at which most minors are treated as adults (Rosch et al. 2007). North Carolina, Connecticut and New York draw the line of adulthood for criminal culpability even at the age of 16. Hence, most research on specific deterrence is based on transferring 16- and 17-year-olds to adult criminal courts such that the validity on the deterrence hypothesis is limited to younger age groups. The German criminal law system provides some interesting alternative institutional framework (see Dünkel, 2005, for historical, institutional and both adult and juvenile law details). Unlike the U.S. system and other European countries, all individuals below 18 are strictly treated under juvenile penal law. Turning 18 does not automatically lead to the application of adult criminal law. Young adults, considered and defined as ‘adolescents’ (‘Heranwachsende’) in terms of the law (18 to under 21 years of age), might be treated the same as juveniles (14 to 18 years old). Whether adolescents were acting ‘[...] equal to a juvenile regarding moral and mental development at the time of the act’ in the sense of § 105 of the German Juvenile Justice Act (‘*Jugendgerichtsgesetz*’, JGG) or not depends on the

discretionary margin of prosecutors and judges.² Similar regulations exist in some other European countries such as Austria, Croatia, Greece, Italy, the Netherlands and Spain (Dünkel 2005).³

This paper contributes to the literature on specific deterrence by addressing the issue of selecting adolescents into adult and juvenile law systems. In particular, we focus on prison inmates belonging to the discretionary ‘adolescent’ group (18 to under 21 years of age) and compare expected recidivism of those convicted under adult criminal law to recidivism of those treated as juveniles. This setting provides new evidence because the design is different from the U.S. paradigm prevailing in the literature. Results are also of related interest for public policy, in particular in Europe, as the German model is used as a fighting crime strategy in other countries.

The data is based on a unique survey on 31 German prisons (German Inmate Survey, see Entorf et al., 2008, Entorf 2009). The econometric approach applies recursive two-equation models (bivariate probit, treatment regressions) which distinguish between a selection equation (selecting individuals to either the adult or the juvenile law) and a treatment equation. The latter captures the causal effect of interest, i.e. the impact of adult treatment on expected recidivism.

Contrary to the bulk of results based on U.S. data found in the criminological literature, results indicate that application of harsher adult criminal laws decreases the expected recidivism of adolescents.

This paper is organized as follows. Chapter 2 gives a short review of related literature. In Chapter 3, a description of the data is provided. Methods and identification strategy are presented in Chapter 4. Chapter 5 informs about econometric results. Chapter 6 gives a short summary and concludes.

² Note that all cases of young adults (<21) are transferred to the jurisdiction of juvenile courts who decide on application of adult or juvenile law.

³ In most European countries, the age of full criminal responsibility is below 21. For instance, application of adult law at the age of 18 is mandatory in Denmark, England/Wales, Finland, France, Sweden and Norway (Dünkel 2005).

2 Previous Research

Research addresses the role of “specific deterrence” and “general deterrence” on future criminal behavior. Rational choice models (Becker 1968) of general deterrence predict that increasing severity and certainty of expected punishments should lead to a decline in crime. Hence, the threat of harsher “adult punishments” will give adolescents pause before they continue committing crimes. Put differently, adolescents have little incentive to curb criminal behavior if they can expect to be treated under the more lenient juvenile law. Evidence in favor of the general deterrence theory was given by Levitt (1998), who showed that the greater the difference between the stricter adult system and the lenient juvenile system, the more the crime rate declined when an increasing number of juveniles were transferred to adult criminal courts. In a more recent paper, Lee and McCrary (2009) found evidence that young adults did hardly respond to the harsher adult law when turning 18, a result that contradicts Levitt’s (1998) findings who showed that there is less crime when adolescents pass into the adult system. In line with the majority of criminologists (see Redding 2003, and Fagan 2008, for surveys), Lee and McCrary (2009) argue that those young offenders can be characterized as myopic, who are unfazed by the threat of short prison sentences and discount the consequences and likelihood of longer imprisonments.

Research on specific deterrence concerns the pros and cons of applying harsher punishment on adolescent offenders in order to reduce the likelihood of future reoffending. Fagan (2008) summarizes the recent criminological literature which is mostly based on U.S. data. The studies typically compare court outcomes and recidivism rates for matched groups of transferred and retained youths. According to his survey, research on transfer of juveniles to adult criminal courts suggests that recidivism rates are either the same or significantly higher for transferred youths than for youths retained in the juvenile court. Some studies compare the criminal records of similar groups of youths either for the same time period or for different time periods before and after law changes. However, as pointed out by Fagan (2008), most of the studies introduce selection biases that prevent a true comparison of individuals with and without the ‘adult’ treatment. A major problem is that many studies do not control for pre-existing indices for criminal propensity that may then affect the outcomes. Hence, differences in the recidivism rates of both samples may reflect more about such pre-existing propensities than about the differential treatment effects. However, some few studies address these selection issues. Winner et al. (1997) matched cases in the juvenile and adult courts on seven

criteria.⁴ Lanza-Kaduce, Lane, Bishop, and Frazier (2005) used propensity scores (Rosenbaum and Rubin 1984) to adjust for selection effects in the transfer process. The differences in recidivism rates using this strategy show substantially higher recidivism rates for transferred youth. Applying the same method, their result is confirmed by Fagan et al. (2004) who compared recidivism rates among samples of youths recruited from New York City whose cases originated in the criminal court, with samples from bordering areas in northeastern New Jersey whose cases were processed in the juvenile court. Further studies by Myers (2003), Podkopacz and Feld (1995), and Thornberry, Huizinga and Loeber (2004) come to the same conclusion.

Hjalmarsson (2009) used a different approach for testing the specific deterrence hypothesis. Contrary to the bulk of research, she is testing the deterrence effect of juvenile facilities when the alternative would be sentencing without incarceration (i.e. probation or fine). Applying a regression discontinuity design that is based on sentencing either to custody or to more lenient measures depending on binding guidelines (“sentencing grid”) for Washington state juvenile courts, she identifies the causal effect by comparing the recidivism rate of individuals who are close to the critical sentencing grid on either side of the cutoff, i.e. juveniles without incarceration on the one side of the grid and juveniles with incarceration on the other side of the grid. Her study shows that incarceration in juvenile facilities can be an effective measure of combating juvenile crime.

A major drawback of the existing empirical studies is the limitation to U.S. data such that most research on specific deterrence is based on transferring 16- and 17-year-olds to adult criminal courts. As outlined in the Introduction, depending on the severity of the crime and the maturity of the defendant, adolescent delinquents can be sentenced according to either juvenile or criminal law. There is no special court for juveniles. Inspecting recidivism rates for those who are convicted either under juvenile law or under criminal law in Germany and released when they were between 18 and 24 years old, it turns out that recidivism rates are higher for those who are treated as juveniles.⁵ Of course, apart from age this comparison

⁴ However, besides information on prior offenses, data on the individual background is rather limited: only age at the time of committing the offense, gender and race are considered.

⁵ Jehle, Heinz and Sutterer (2003) analysed official register data on recidivism within a four years period (1994 to 1998). The recidivism rate after being released from an unconditional prison sentence under juvenile law was 79.0% for 18-to-20-year-olds and 74.9% for 21-to-24-year-olds (Jehle et al. 2003, p. 109). Respective recidivism rates after unconditional prison sentences under adult criminal law were 46.2% (18 to 20 years of age) and 49.9% (21 to 24) (Jehle et al. 2003, p. 107). Aggregating over all age groups, the recidivism rate within four years after unconditional prison sentence under juvenile law was 77.8%, whereas it was 54.9% for adults.

might still be biased by selection due to other factors such as education, family background or labor market prospects (Entorf 2009).

3 Data and Descriptive Evidence

This analysis is based on a prison survey of 1,771 German inmates. The survey (German Inmate Survey, Entorf et al., 2008) uses a two-stage approach that combines stratified and random sampling. First, a representative sample of the population of prisons in Germany was created. Second, a random draw out of this population completed the sampling.

The survey data was collected in 31 German prisons in 2003 and 2004 using a questionnaire with 123 questions. In order to account for the different nationalities of the inmates, the questionnaire was distributed in either German, Turkish, Serbo-Croatian, Russian, Polish or English language. The survey was organized and performed by a team of researchers from Darmstadt University of Technology. It was first analyzed by Entorf (2009).⁶

The convicts of the dataset can be grouped into three subsamples: inmates in pretrial custody, inmates sentenced under juvenile law and inmates sentenced under criminal law. Since this study focuses on the effect of the applied law type, only the last two subgroups are used and inmates in pretrial custody are dropped. Furthermore, as the main research interest is on juveniles and young adults, only individuals aged between 15 and 22 when they committed the crime are considered relevant for the analysis. This results in a subsample of 159 inmates. The descriptive statistics can be found in Table 1.

The crucial variable of interest is expected recidivism. It is constructed from the response to the following survey question:

(1) Could it occur that after your release from custody you come into conflict with the law and end up in prison?				
Tick your assessment on the following 5-point scale, whereby a 1 stands for “no, never”, and a 5 stands for “absolutely certain”:				
No, never				Absolutely certain I’ll be in again
<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
1	2	3	4	5

⁶ Interviews with judicial employees suggest an overrepresentation of the more active group of inmates. However, when comparing the sample characteristics to those of the average prison population in Germany, there no evidence of a selection bias was found (see Entorf et al. 2008).

Answers to this question are transformed into a binary variable *expected recidivism*. Given the ordered categories of the survey question, ordered discrete choice models would have been the natural alternative to the performed binary response model. The reason for summarizing categories is the rather small number of 159 observations in the age-at-offense group of interest. Summarizing categories prevents misinterpretations of subjective survey questions and avoids problems arising from sparsely filling in certain survey categories, in particular due to the often found rare use of extreme survey categories.

Table 1: Descriptive Statistics, Conditional on *age at offence* $\in (15, 22]$

Variable	Nobs	Mean	Std. Dev.	Min	Max
recidivism (expectation)	159	.2893082	.4548739	0	1
age at offense	159	18.9908	1.870697	15.25	22
age	159	21.18868	2.679434	16	29
male	159	.9308176	.2545658	0	1
married	159	.081761	.274866	0	1
poor social capital	159	.4842767	.5013317	0	1
no church	159	.3396226	.4750775	0	1
criminal family background	159	.2201258	.4156404	0	1
Abitur	159	.0503145	.2192837	0	1
addiction (alcohol or drugs)	159	.2893082	.4548739	0	1
job contact	159	.1949686	.3974282	0	1
schooling	159	9.836478	2.102212	2	17
length (of prison sentence)	157	3.167994	2.641088	.167	15
(application of) criminal law	159	.3018868	.4605269	0	1
murder	159	.1194969	.3253971	0	1
sexual offence	159	.0503145	.2192837	0	1
assault	159	.3144654	.4657696	0	1
robbery	159	.3117647	.4645827	0	1
theft	159	.4150943	.4942951	0	1
fraud	159	.1823899	.3873857	0	1
illicit drugs: consumption	159	.0754717	.2649855	0	1
illicit drugs: dealing	159	.1383648	.3463734	0	1
vandalism	159	.1194969	.3253971	0	1

Note: descriptive statistics under the restriction that variables show valid observations on ‘recidivism’; Data Source: German Inmate Survey (Entorf et al. 2008).

In the data, the answers are positively skewed: 43.5% of the respondents answered with the lower extreme “No, never” while only 4% said they were absolutely certain to reoffend. Therefore, we set expected recidivism to zero if the respondent chose either answer 1 or 2, and set the binary variable to one for those with a higher self-reported probability of ending up in prison again (answers 3-5).

One might raise objections against using perceived expected recidivism as a proxy for real recidivism and actual criminal behavior. However, there are at least three arguments in favor

of self-assessments. First, there is evidence that expected and real recidivism are correlated (see for instance Corrado 2003, Lochner 2007). Second, using expected recidivism in contrast to actual recidivism avoids the problem of a selection bias which arises when conducting follow-up surveys collecting information on realized recidivism. Third, when studying joint distributions of expected recidivism and related covariates researchers do not have to rely solely on unconditional recidivism rates (without any meaningful background information such as socioeconomic or family situations) as they can be found in most recidivism surveys (as in Jehle et al. 2003, for German recidivism rates).

The crucial variable for the assigned type of law is the age at the time when the crime was committed (*'age at offense'*). Since this information did not appear directly in the survey, it is constructed by using time and age when inmates were interviewed and the time when the crime was committed. With regard to the latter, inmates could choose to indicate either a point in time or an interval. In the case of a point in time, the calculation is straightforward. When dealing with an interval, it is not very clear which point in time to take. For these cases, we took the upper bound of the interval, since German juvenile law (*Jugendgerichtsgesetz, JGG*) allows judges to apply adult criminal law once the delinquent has reached the critical age when committing his criminal acts (§ 32 JGG).

Table 2 informs about expected recidivism of juveniles (*age at offense < 18*) dependent on *age* (at the time of the interview). Young offenders who committed their crimes when they were younger than 18 seem to be very pessimistic when still young (below 18): Here, expected recidivism is 0.5 which is clearly above the average rate of 0.29. Older cohorts originating with the same *age at offense* show less pessimistic assessments of their future, and after a couple of years (*age ≥ 21*) they even seem to respond in a rather optimistic way (0.14), although this evidence cannot be generalized because it is based on only 7 observations.

Table 2: Expected Recidivism of Juveniles (*age at offense < 18*)

Age (at time of survey)		
<18	[18, 21)	≥ 21
0.50 (8)	0.29 (41)	0.14 (7)

Notes: Share of inmates with “expected recidivism = 1” whose age at the time of the offense was below 18; number of respondents in parentheses. Data: German Inmate Survey (Entorf et al., 2008).

The age group [18,21) is of particular importance because for criminals of this age German judges decide about application of either criminal or juvenile law (JGG). Table 3 considers expected recidivism of the age-of-offense group [18,21), and it distinguishes between those who are convicted under juvenile law and those who are sentenced under adult conditions. Preliminary descriptive evidence suggests that those convicted under juvenile law expect markedly higher rates of recidivism, conditional on the age at the time of the interview. Ignoring age, the average recidivism rate of ‘juveniles’ would only be slightly higher than that of ‘adults’ (0.34 compared to 0.31).

Table 3: Expected Recidivism of Adolescent Offenders

Age at time of offense (related to present prison sentence) = [18, 21)		
Age (at time of interview)	Application of juvenile penal law	Application of (adult) criminal law
[18, 21)	0.35 (20)	0.00 (1)
[21, 23)	0.29 (31)	0.25 (8)
≥ 23	0.60 (5)	0.35 (17)
Average (all age groups)	0.34 (56)	0.31 (26)

Notes: Share of inmates with “expected recidivism = 1” whose age at the time of the offense was at least 18 but under 21 years of age; number of respondents in parentheses. Data: German Inmate Survey (Entorf et al. 2008).

Of course, *age* and maturity of young criminals are important factors of recidivism, but many other covariates are of similar relevance. One of these variables is gender (*male*). Consistent with national criminal survey statistics, there is a strong majority (90%) of male inmates in our sample. We also control for marital status (*married*). In our sample, only 8.2% are married, which is probably due to the fact that we are looking at inmates aged 15 to 22 when committing the crime. Further, we measure participation in social clubs, e. g. sports clubs or the voluntary fire brigade, mapping the lack of active participation into the dummy variable *poor social capital*. Almost half of the inmates in question (48.4%) reported no active participation in social clubs. Furthermore, religious affiliation might play an important role in terms of beliefs and intrinsic motivation. The corresponding dummy variable *no church* is set to one for all inmates that do not belong to any religion which holds true for 33.9% in our sample.

Crime and education are often found to be negatively correlated. This study includes education measured as the number of years spent in school (*schooling*). As education is a positive asset in the legal labor market, it should reduce the probability to reoffend. However,

education might also be of importance to judges who have to decide to what extent suspects are fully responsible for their behavior or whether individual maturity is lagging behind. Poor or high educational levels of adolescents might help judges in making her or his decision about which penal law to choose. *Abitur* measures the educational degree which is usually achieved after 18 or 19 years of age; it qualifies for university entrance in Germany. For the relevant subgroup of inmates with *age at offense* between 18 and 21 years the ratio is 5.3%. (full sample: 5.0%, see Table 1).

Recidivism heavily depends on future legal perspectives. Labor market opportunities play a leading role. The analysis includes the binary variable *job contact* which is set to one for all inmates reporting to have a job opportunity when leaving jail which is the case for 19.5% of the relevant inmates.⁷ Recidivism is likewise linked to alcohol and drug *addiction*, both in a direct way – in the sense that addicted people might more easily commit crimes under the influence of drugs – and in an indirect way in the sense that these people might see criminal behavior as a way to finance their habits (see for instance Entorf & Winker 2008, Goldstein 1985 or Harrison 1992). In the sample, 30.2% of the inmates suffer from at least one of these addictions. Furthermore, *criminal family background* could matter for expected recidivism. This dummy variable captures past convictions of parents or siblings and applies for 22% of the inmates in the sample.

Finally, the survey contains detailed information on the crime committed and the corresponding sanction. Many delinquents have a criminal record of different offences resulting in multiple crime types being reported by a single individual. The most frequently reported crime is *theft* (41.5%), followed by *assault* (31.4%) and *robbery* (31.2%).

In terms of applied legislation, 30.2% of all delinquents in the sample were sanctioned under adult *criminal law*. For the important subsample of those who committed their crime between 18 and 21 years of age, this share is 28.2% (Table A3, Appendix). We also observe the *length* of a prison sentence measured in years. In line with German legislation, lifelong punishments (recorded as ‘lifelong’ in the data) are recoded as a 15 year sentence, which is also the maximum custody length found in the sample.

As mentioned above, depending on *age at offense* different laws might apply. Delinquents can thus be grouped into three groups. Juveniles can only be sentenced according to juvenile law (Table A2, 55 observations), adolescents can be penalized by either juvenile or criminal law

⁷ See Entorf (2009) for a discussion of the influences of school and job market opportunities on criminal behavior and recidivism.

(Table A3, 78 observations) and adults who have to be sentenced under criminal law (Table A4, 26 observations). Thus, to sharpen the contrast between juvenile law and adult law, the sample of the fuzzy age-at-offence group [18,21) is enhanced by criminals of similar age either strictly subject to adult or strictly subject to juvenile law. Subsequent econometric results will be based on both samples, i.e. age at offence $\in (15,22]$ and $\in [18,21)$.

A comparison of the three groups reveals some significant differences at the descriptive level.⁸ There are only male juveniles and more than 93.6% male adolescents while we have 23.1% female adult offenders in our sample, representing a statistically significant drop. As expected, there is a statistically significant rise in the share of married respondents when moving from juvenile to adolescent offenders. Furthermore, the years of schooling increase in age yielding a significant difference between juveniles and adolescents.

In addition, there are further notable but non-significant differences between groups. The rate of *recidivism* seems to be similar in the first two groups (approx. 30%) but lower for the group of adults (18%). Also, job contacts decrease with age which seems to indicate that inmates have a harder time to reenter the legal job market when getting older. We find sentence length to increase with age, which seems natural because adult persons get more severe sentences. In terms of crime types we see that almost all crimes become less frequent as the age increases. Especially interesting is vandalism, which does not show up at all for adult (>21) delinquents. The only crimes that are increasing with age are fraud, sexual offences, drug dealing and drug consumption. However, as for all differences described in this paragraph, the change between the groups is not significant.

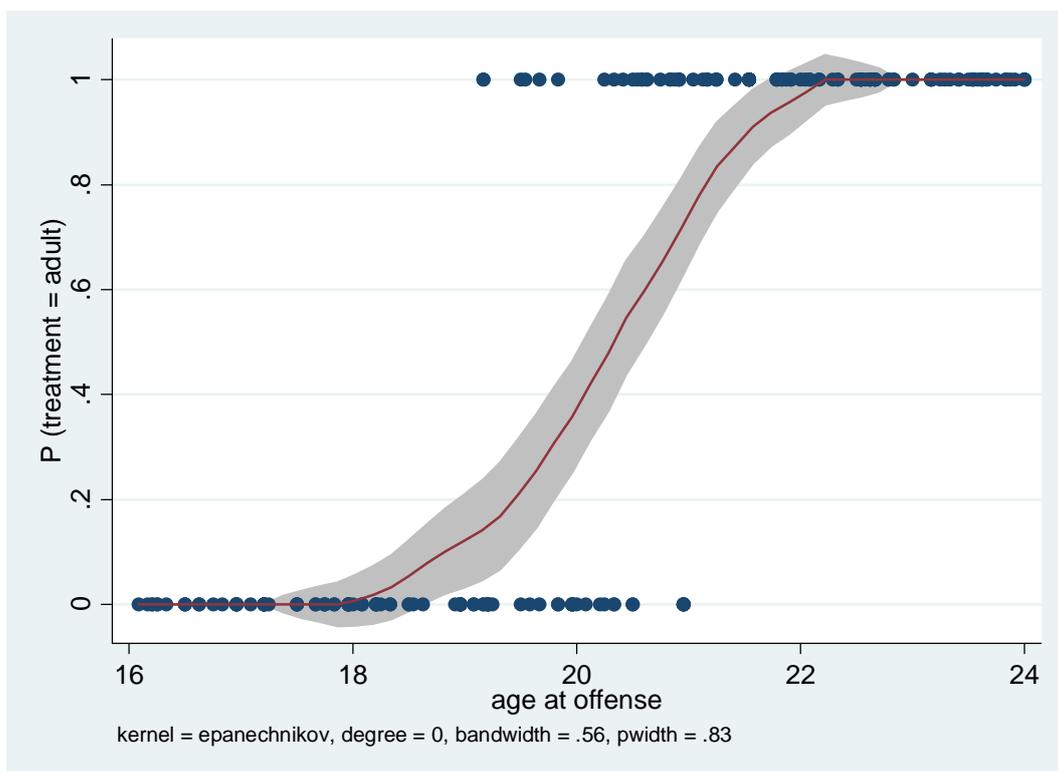
4 Legal Background and Econometric Identification Strategy

The main goal of this study is to focus on young adults ('adolescents') who were at least 18 years of age when they committed their offences and to compare the expected recidivism of those who are *sentenced under adult criminal law* ('*Strafgesetzbuch*', StGB) to the one of adolescents sentenced *under juvenile penal law* ('*Jugendgerichtsgesetz*', JGG). According to German legislation, juvenile law is mandatory for all delinquents aged less than 18 when committing the crime. For offenders of at least 18 but less than 21 years of *age at offense*, the legislator (§ 105 JGG, *Jugendgerichtsgesetz*) left the decision to the (juvenile) courts whether

⁸ In terms of the difference of two sample means we consider them as statistically significantly different from each other using a two sample t test if the significance of their difference is higher than 10%.

to apply juvenile or criminal law. Finally, delinquents of at least 21 years have to be sentenced under general criminal law. With regard to the treatment assignment function, German legislation specifies that criminal law should be applied whenever the adolescent offender does not appear to be similar to a juvenile, neither in terms of personal development nor in the type of crime committed. Using these maturity criteria, judges of adolescents of the age-at-offense group [18, 21) have to evaluate which law to apply. It is quite obvious that the judge's decision will hinge on observables (like age) but also on unobservables (like maturity) which, however, likewise determine expected recidivism. Thus, when evaluating the recidivism rate of both penal law regimes, there is a potential endogeneity problem and the treatment effect cannot be estimated by simply inspecting sign and magnitude of treatment dummy variables.

Figure 1: Estimated Probability of Applying Adult Criminal Law



Note: Local Polynomial Regression, incl. 95% confidence interval; Data: See Table 1.

The first idea was to tackle the endogeneity problem by using the regression discontinuity design (see Imbens and Lemieux 2008, and Van der Klaauw 2008, for surveys), and by inspecting recidivism rates just before and after turning 18. However, the propensity score is almost zero until 'age at offense' equals 19 years of age at the time of committing the offense,

as can be seen from *Figure 1*.⁹ Thus, RD does not work smoothly, neither in a sharp nor a fuzzy framework, when the legislative cut-off point ‘age at offense=18’ is not effective, even not at some closer neighbourhood. Entorf et al. (2010) experimented with different alternative cut-off points at *age at offense* > 18 which, however, clearly bears the major problem of considerable arbitrariness.

In order to transfer the legal restrictions into a statistical model, the framework has to deal with the problem that turning 18 entails a change from definite application of juvenile penal law (applied to all minors under 18 years of age) to a different ‘fuzzy’ law status. The fact that juveniles turn 18 might already entail some deterrent effect on its own, irrespective of being treated as juvenile or adult. At the same time, turning 18 implies a second (nested) and potentially more severe shift from juvenile law to adult criminal law, given the person is convicted accordingly. Thus, the equation of interest on expected recidivism should provide information on the effect of both cutoffs, i.e. ‘being at least 18 years of age or older at the time of committing the offence’ and ‘being at least 18 years of age at the time of committing the offence *and* being convicted under criminal law’. An equation that accounts for both effects on expected recidivism of inmate *i* (in prison *j*) is

$$(1) \quad ER_i = z_i\beta + \rho_1 D_i + \rho_2 T_i + \varepsilon_i,$$

where $D_i = 1$ if the person was at least 18 years old at the time of the offence (0 otherwise), and where the treatment $T_i \equiv D_i A_i = 1$ indicates application of the adult criminal law with $A_i = 1$ representing adult criminal law, $A_i = 0$ juvenile penal law.

Equation (1) includes a vector of explanatory factors z_i and a residual ε_i . The latter might include some prison specific influences which are taken into account by correcting for unobserved clustering effects. Obviously, the interpretation of the treatment effect ρ_2 would be limited when we estimated eq. (1) by ordinary least squares because of the selection problem outlined before: Young inmates who committed their crimes when they were above the critical age of 18 (but still younger than 21) are committed under ‘adult’ conditions for good reasons, which might include some more serious criminal backgrounds, advanced criminal careers and a higher risk of recidivism (when convicted as juvenile). Thus, treatment as an

⁹ Actually, for those inmates who are treated under adult conditions the youngest age at the time of committing the offence found in the survey is 19. The local polynomial regression (using optimized bandwidths) provides smooth transitions from zero to non-zero probabilities.

adult, T_i , should be considered as endogenous, depending on correlated and common factors which motivate courts treating young adults under (adult) criminal law:

$$(2) \quad T_i^* = x_i\gamma + e_i, \quad \text{where } T_i = \begin{cases} 1 & \text{if } T_i^* > 0 \\ 0 & \text{otherwise} \end{cases} \quad \text{and } \text{cor}(\varepsilon_i, e_i | z, x) = r.$$

Under the hypothesis that being treated as adult or not depends on both observable and unobservable factors covering the propensity for continuing the criminal behavior, it is assumed that unobserved factors which cause judges to rule under adult law lead to higher recidivism of convicted offenders: Thus, $\text{cor}(\varepsilon_i, e_i) = r$ is expected to be positive.

The econometric problem is that both recidivism in eq. (1) and the potentially endogenous ‘adult’ treatment in eq. (2) are binary variables. A straightforward approach would be to estimate eq. (1) by applying a two-stage probit procedure (à la 2SLS), i.e. first run a probit of T on all exogenous variables and get the fitted probability $\hat{\Phi}$, then run probit on z, D and $\hat{\Phi}$. However, as pointed out in the literature (see Imbens and Wooldridge (2009, lecture 6, Angrist and Pischke 2009, Chapter 4) estimating these so called “forbidden regressions” does not work because it would require passing the expected value through a nonlinear function. As there are no known two-step methods that allow one to estimate probit models with discrete endogenous variables, Imbens and Wooldridge (2009, lecture 6, p. 28) advocate the use of bivariate probit models with binary endogenous explanatory variables. Accordingly, a bivariate probit model in the tradition of Maddala (1983, p.123) will be estimated (it is called a ‘recursive, bivariate probit model’ by Greene and Henscher 2010, p. 90):

$$(3) \quad \begin{aligned} P(T = 1 | x) &= \Phi(x\gamma) \\ P(ER = 1, T = 1 | z, D, T, x) &= \Phi_2(z\beta + \rho_1 D + \rho_2 T, x\gamma, r). \end{aligned}$$

Note that the endogenous nature of T on the right-hand side variable of the second equation can be ignored in nonlinear maximum likelihood estimations (see, e.g., Greene 2008, p 823, Imbens and Wooldridge, lecture 6, p. 28). Another advantage of the bivariate probit model compared to the linear case is that there is no exclusion restriction required for identification: exogenous variables may appear in both equations (Greene and Hensher, 2010, p. 91). The only restriction is that there needs to be at least one exclusion restriction in the bivariate probit first stage (Angrist and Pischke 2009, p. 199).

Primary interest is in the consistent estimation of the parameter ρ_2 (and not in estimating endogenous probabilities of recidivism) such that treatment-effects models can be considered

as alternative estimation strategy. They test the effect of an endogenously chosen binary treatment on another endogenous continuous variable, conditional on two sets of independent variables. Thus, unlike in bivariate probit models, endogenous expected recidivism will be estimated within a linear probability model.¹⁰ Details are outlined in Maddala (1983, 117-122) and Greene (2008, p. 889-890). The model makes use of Heckman's (1979) correction approach (for endogeneity, not selection) such that the expected rate of recidivism for those treated under adult criminal law can be estimated using the inverse Mills ratio:

$$(4) \quad \begin{aligned} E(ER_i | T_i = 1, x_i, z_i) &= z_i\beta + \rho_1 + \rho_2 + E(\varepsilon_i | e_i > -x_i\gamma) \\ &= z_i\beta + \rho_1 + \rho_2 + r\sigma_\varepsilon \frac{\phi(x_i\gamma)}{\Phi(x_i\gamma)}. \end{aligned}$$

The respective rate for those treated under juvenile law is

$$(5) \quad \begin{aligned} E(ER_i | T_i = 0, x_i, z_i) &= z_i\beta + \rho_1 + E(\varepsilon_i | e_i < -x_i\gamma) \\ &= z_i\beta + \rho_1 + r\sigma_\varepsilon \left[\frac{-\phi(x_i\gamma)}{1 - \Phi(x_i\gamma)} \right]. \end{aligned}$$

Identification of the causal treatment effect of interest, ρ_2 , requires that exogenous factors of selectivity are included in x_i and to have variables in z_i that are not included in x_i . This exclusion restriction will be satisfied in the subsequent analysis. As already discussed in the previous section, courts consider age as a major factor influencing their decision. Moreover, types of crime and indicators of maturity will be used as explanatory variables in the selectivity equation. Individual labor market opportunities, social capital as well as alcohol and illicit drug problems serve as main factors z_i of recidivism in the equation of interest.

5 Results

Table 4 presents results for the bivariate probit and treatment-effects models outlined above. Both frameworks combine the equation of interest (estimating the causal effect of the adult criminal law treatment) with a second equation which maps decisions of courts. They select adolescents/young adults either to the adult or the juvenile regime. All results are based on maximum-likelihood estimations (or, as argued in Huber 1967, and White 1982, on quasi-

¹⁰ Angrist (2001) promote just using linear probability models estimated by 2SLS. The Heckman-two step model can be interpreted as a related control function approach. For conditions leading to equivalence of 2SLS and control functions see Imbens and Wooldridge (2009, lecture 6).

maximum-likelihood if the distributional assumptions are not fully correct). The calculation of standard errors considers potential effects arising from stratification and clustering (in prisons). The sample is restricted to young prison inmates, because the focus of the study is on the distinction between juvenile law applied to defendants younger than 18 years of age and a case-based discretionary law applied to adolescents between 18 and 21 years of age. Empirical evidence relies on two nested age groups. The first one includes inmates of neighbouring age groups whose age at the time of the offense was between 15 and 22, the second one is strictly limited to the group of 18 to 21 years of (offender) age.

Before turning to the results of two-equation models in more detail, some general observations can be reported. As endorsed by estimated single-equation probit models (Table 4 column (1)) and subsequent two-equation models, *expected recidivism* is higher for those with *criminal family background* (i.e. parents or sisters have a criminal record), *drug or alcohol addiction* and for inmates with *poor social capital*. Recidivism seems to be less likely when a *job contact* or more years of *schooling* lead to more optimistic labor market perspectives. These results confirm previous probit results based on all survey inmates (1,568 observations) presented in Entorf (2009). The only exception is that social capital was not significant in the full sample whereas it is in the subsample. The strong and highly significant effect throughout all specifications for the young cohort shows that lack of social capital is particularly hazardous for juvenile and adolescent individuals.

Some variables were found insignificant for *expected recidivism* in column (1) (Table 4) as well as in more complex two-equation models such that they were disregarded in further models of Table 4. These are *no church* (indicates whether the minor is a member of a protestant, catholic, islamic, etc. church or not), *male* and *age at offense*. The insignificance of gender (found throughout all models) is somewhat surprising as results based on the full sample (Entorf 2009) suggest markedly more optimistic future perspectives for females. There might be two explanations for the less optimistic prospects of the young female prison cohort under consideration. The first one is that adolescent females in prisons represent an adverse selection, even regarding the universe of all females and young males in prison, possibly because judges apply the *ultima ratio* of unconditional imprisonment for young females in a more conservative way than they do for young males or more mature females. The second explanation is that there is convergence of female and male behavior or attitudes such that assessments made by young female cohorts cannot be distinguished any more from those made by their male counterparts.

Table 4: Econometric Estimation of Two-equation Models

	Expected Recidivism					
	(1) Single Probit 15<x ≤22	Part I of two-equation models				
		(2) Bivariate Probit 15<x ≤22	(3) Bivariate Probit 15<x ≤22	(4) Bivariate Probit 18<x ≤21	(5) Treatment Regression 15<x ≤22	(6) Treatment Regression 18<x ≤21
$D (x \geq 18) (\rho_1)$	0.650 (0.378)	0.414 ^{***} (0.150)	0.384 ^{**} (0.145)	–	0.121 ^{**} (0.052)	–
$T (= adult) (\rho_2)$	-0.576 (0.644)	-1.573 ^{**} (0.731)	-1.487 ^{**} (0.661)	-1.604 ^{**} (0.612)	-0.459 ^{**} (0.207)	-0.530 [*] (0.253)
Drug or alcohol addiction	0.477 ^{**} (0.199)	0.371 [*] (0.197)	0.410 [*] (0.207)	0.617 [*] (0.309)	0.144 [*] (0.071)	0.207 [*] (0.113)
Criminal family background	0.495 ^{**} (0.188)	0.465 ^{**} (0.172)	0.422 ^{**} (0.191)	0.578 ^{**} (0.272)	0.146 ^{**} (0.059)	0.168 (0.105)
Job contact	-0.341 [*] (0.174)	-0.460 ^{***} (0.133)	-0.386 ^{***} (0.127)	-0.643 (0.428)	-0.108 ^{***} (0.033)	-0.146 (0.106)
Poor social capital	0.554 ^{***} (0.166)	0.635 ^{***} (0.175)	0.566 ^{***} (0.176)	0.802 ^{***} (0.267)	0.166 ^{***} (0.054)	0.197 ^{***} (0.057)
No church	0.110 (0.147)	0.101 (0.132)	–	–	–	–
Male	-0.058 (0.398)	–	–	–	–	–
Schooling	-0.102 [*] (0.054)	-0.089 [*] (0.051)	-0.089 [*] (0.048)	-0.156 [*] (0.078)	–	–
Age < 18	-0.600 (0.717)	-0.471 (0.622)	-0.453 (0.606)	–	-0.102 (0.185)	–
18 ≤ age < 21	-0.837 (0.651)	-0.943 (0.680)	-0.905 (0.663)	-1.163 [*] (0.661)	-0.265 (0.210)	-0.361 (0.258)
21 ≤ age < 22	-0.874 (0.516)	-0.944 [*] (0.472)	-0.991 [*] (0.483)	-0.935 [*] (0.497)	-0.311 [*] (0.164)	-0.373 [*] (0.208)
Age at offense (‘x’)	-0.219 (0.183)	–	–	–	–	–
r [P(LR of r = 0)]	–	0.764 ^{**} [0.023]	0.698 ^{**} [0.032]	0.744 [*] [0.054]	0.488 ^{**} [0.030]	0.532 ^{**} [0.046]
$r\sigma_\varepsilon$	–	–	–	–	0.211 ^{**} (0.071)	0.236 [*] (0.096)
P(LR-Test)	0.035	0.000	0.000	0.000	0.004	0.009
Nobs.	159	159	159	78	159	78

Table 4: (Continued)

	Conviction under Adult Criminal Law							
	(1') Single Probit	Part II of two-equation models					(6)	
		(2) Bivariate Probit	(3) Bivariate Probit	(4) Bivariate Probit	(5) Treatment Regression	Treatment Regression	Marginal effects (on recidivism)	
Criminal family background	0.585* (0.308)	0.330 (0.263)	–	0.596* (0.283)	–	–	–	
Poor social capital	0.589 (0.383)	0.603 (0.385)	–	0.641* (0.349)	–	–	–	
No church	0.428 (0.454)	0.374 (0.448)	–	–	–	–	–	
Schooling	–	0.056 (0.055)	–	–	–	–	–	
Abitur	1.584** (0.583)	–	1.398** (0.619)	1.405** (0.522)	1.433** (0.623)	1.250* (0.716)	-0.207*** (0.072)	
Age at offense ('x')	1.081*** (0.268)	1.439*** (0.246)	1.419*** (0.250)	1.170*** (0.252)	1.414*** (0.237)	1.144*** (0.246)	-0.213*** (0.082)	
Short sentence (< 2 years)	0.577 (0.438)	0.198 (0.526)	–	–	–	–	–	
Assault	0.040 (0.392)	–	–	–	–	–	–	
Sexual offense	0.073 (0.909)	–	–	–	–	–	–	
Theft	-0.342 (0.574)	-0.750* (0.431)	-0.487 (0.355)	-0.453 (0.510)	-0.461 (0.384)	-0.404 (0.500)	0.076 (0.106)	
Robbery	-0.868 (0.533)	-0.957** (0.348)	-0.753** (0.314)	-0.867** (0.367)	-0.723** (0.319)	-0.746* (0.361)	0.142** (0.064)	
Illicit drug use	-0.812 (0.641)	-0.621 (0.586)	–	–	–	–	–	
P(LR-Test)	0.000	(see equation on expected recidivism)						
Nobs.	78							

Notes: Joint ML estimates of two-equation models (1) and (2). Part II represents eq. (2). Clustered standard errors are shown in parentheses. P(LR-Test) indicates the p-value of the likelihood-ratio test of the hypothesis $r=0$. Marginal Effects of binary variables w are calculated as $E(ER | T = 1, w = 1, x, z) - E(ER | T = 1, w = 0, x, z)$; corresponding standard errors are obtained using the delta method. Data: See Tables 1 and A3.

Age at offense is expected to be insignificant because this variable is not directly related to expected recidivism when *age* of inmates (at the time of the interview) is already controlled for. As descriptive statistics revealed some nonlinear pattern of the influence of age (see also Table 2, Table 3), it is captured using dummies for age intervals. Inmates included in the regression were between 16 and 29 years old when they were interviewed (see Table 1). In Table 4, the (omitted) reference group is 22 years of age or older. There seem to be age-dependent moods of inmates which can be characterized as increasingly optimistic until the age of 21 and more pessimistic when getting older than 21 years of age. In order to exclude potential age-dependent nuisance, age dummies are retained throughout all specifications despite their lack of strong statistical significance.

The idea behind the assignment of explanatory variables to the selection and the treatment equations is to discriminate between the individual personal situation at the time of committing the offense (reconstructing the court's information), and the situation prevailing at the time of the interview. Among the relevant data leading to informed decisions of judges, the type of crime, the degree of maturity and the social situation of defendants should play a role. Turning to estimated equations analyzing the probability of being treated under adult criminal law (Table 4, panel II), results are strongly in line with Figure 1 and confirm the crucial role of age at the moment of committing the crime in question (*age at offense*). Moreover, a higher degree of education, 'Abitur'¹¹, increases the probability of being treated as adult.¹² These results suggest that judges not only consider age but also intellectual maturity as indicators of culpability in the sense of the adult criminal law. Besides these factors, criminal family background and poor social capital seem to have some (weak) relevance for deciding in favor of adult criminal law, possibly because both variables are positively related to unobserved factors such as lack of moral norms, lack of insight in wrongdoing or lack of compliance - all reflecting judges' lack of confidence in future law-abiding behavior. As regards types of crimes, robbery is clearly treated under juvenile crime. Thieves, too, have a higher probability of being convicted as a juvenile, but to a lesser and mostly insignificant extent. *Vandalism* had to be excluded from the analysis, because, with few exceptions, all cases were treated under juvenile law such that probit achieved perfect predictions of endogenous outcomes. In Table 4, the most severe crimes, *murder* and *sexual offense*, are not included in most specifications such that these categories always belong to the reference group (note, however, that inmates can serve for more than one crime). The same holds for *drug dealing*. Several specifications experimenting with the severity of sanctions show no significant result. Table 4, column (1) presents a model that includes the impact of 'short' prison terms which is found to be positive (but insignificant). As already observed for expected recidivism, religious affiliation has no significant effect on being treated either as adult or as juvenile.

The main focus is on the identification of the causal effect of treatment under 'adult' conditions. The structural parameters of interest capturing the threshold effect between juvenile and non-juvenile law (ρ_1) and the effect of adult criminal law (ρ_2) are both

¹¹ 'Abitur' is the general qualification for university admission in Germany; it is equivalent to the American high-school diploma, the French 'baccalauréat' and the British 'A level'.

¹² Alternative specifications using years of education lead to the same conclusions, but *schooling* turned out to be only weakly significant in most specifications such that the degree 'Abitur' was used instead.

insignificant in single equation probit models. However, the single-equation model of recidivism is affected by some ‘omitted variable bias’ in the sense that the unobserved factor e in the recidivism equation (1) is correlated with unmeasured random determinants captured by ε in the assignment equation (2). As described in detail in the previous Section, ML estimations of two-equation models presented in column (2) to (6) of Table 4 take this correlation into account. All results, irrespective of estimation method and model specification, indicate a high and positive correlation coefficient r . Likelihood-ratio tests show significance at the 5% level (p-values range between 0.023 and 0.054). The positive correlation confirms the presumption that adverse but unobserved common factors such as obvious criminal energy and persistence of unreasonable and unteachable attitudes which lead judges to rule under criminal law cause higher expected recidivism rates.

All two-equation models indicate that the parameter ρ_2 on application of general criminal law is negative and significant at the 5% level (see models (2) to (5) of Table 4) or at the 10% level in model (6). These estimates indicate that application of criminal law instead of juvenile law decreases expected recidivism. This result is in line with the hypothesis that treatment under the stricter adult law entails some deterrent effect such that inmates intend to avoid future imprisonment under adult conditions. The alternative explanation is that prisoner rehabilitation and corrective measures are more effective under adult law than under juvenile law. However, the latter view seems to be rather unlikely given the fact that costs per juvenile prisoner exceed those of adult prisoners by about 40 to 50% which is mainly due to a much higher ratio of personnel per inmate.¹³

The positive estimate of ρ_1 , i.e. the parameter which represents the overall shift from ‘pure’ juvenile law to a discretionary regime of juvenile and adult law, reveals a smaller but significant upward shift of expected recidivism after turning 18.¹⁴ Thus, those being older than 18 years of age seem to be less deterred by the mixed regime than those below the threshold of 18 years of age who are strictly treated as juveniles. Again, two interpretations are in order. The first one is that the threat of a weighted risk of juvenile law and adult law increases the inclination to commit future crimes, possibly because inmates know that their personal risk of being affected by adult law is relatively low and because juvenile custody is

¹³ A survey among the management of German prisons (Entorf et al. 2008, p.144) revealed that the average number of employees per 100 adult inmates amounted to 50 people, whereas the manpower invested for supervision and rehabilitation of 100 juvenile inmates was 71 persons.

¹⁴ Of course, this effect has to be interpreted as the average effect of all ages exceeding 18, i.e. it is different from the local effect one would get from a RD design.

considered not as bad by (criminally experienced) young adults older than 18 years than by ‘true’ juveniles of, say, 16 or 17 years of age. The second interpretation is that rehabilitation measures are less effective for adolescent offenders.

Looking at bivariate probit results in more detail (Table 4, columns (2) to (4)), we observe that estimates do not change much when we reduce the age-at-offence window from (15, 22] to [18,21) (such that ρ_1 cannot be identified). In the recidivism equation, only job contact loses significance, whereas criminal family background and poor social capital even gain significance in the assignment equation.

Table 4, columns (5) and (6), show estimates for the treatment approach. Specifications differ somewhat from the bivariate probit because inclusion of the inverse Mills ratio in the equation of interest requires some stricter consideration of the exclusion restriction. Hence only *age at offence* (as general indicator of maturity), ‘*Abitur*’ (representing intellectual maturity) and type of offence are used as exogenous factors which are supposed to be of crucial importance for selecting adolescents into either juvenile or adult law, whereas the significance of all remaining variables is tested within the treatment equation. Signs and significance levels of the larger sample do not differ from the ones of bivariate probit (compare columns (5) to (3)), whereas for the smaller sample estimates of the treatment regression are somewhat less significant than those of bivariate probit (see columns (4) and (6)). The high significance of the correlation coefficient r is confirmed. Moreover, $r\sigma_\epsilon$ is positive and significant (although only at the 10% level in the smaller sample) such that omitting the inverse Mills ratio from the treatment equation would lead to an omitted variable bias.

The last column in the assignment equation shows marginal effects of exogenous factors on expected recidivism (note that the treatment regression is based on the linear probability models such that estimated parameters are identical to marginal effects). These marginal effects are calculated under the restriction that the selection decision is $T=1$, i.e., e.g., for the dichotomous variable *Abitur* it reads $E(ER | T = 1, Abi = 1, x, z) - E(ER | T = 1, Abi = 0, x, z)$. Marginal effects have opposite signs. This is quite reasonable as can be seen from *age at offence* or *Abitur*: Higher maturity increases the probability of being treated as adult which, however, implies that after being selected to such ‘adult’ conditions adolescents show lower recidivism rates.

6 Summary and Conclusions

This paper addresses the issue of appropriate treatment of young adult offenders. In particular, it is an open question whether they should be treated under juvenile law (with strong emphasis on educational and correctional measures), or under application of the more severe adult criminal law. However, when comparing recidivism rates of both groups, most of the existing surveys studies introduce selection biases that prevent a true comparison of individuals with and without the ‘adult’ treatment. A major problem is that many studies do not control for pre-existing conditions that may then affect the outcomes.

This paper contributes to the literature on specific deterrence in several ways: First, the study based on the German legal system (which is similar to the one in several other European countries) provides some external evidence to the prevailing literature which is based on the U.S. paradigm, where the line of adulthood is at some younger ages. Second, unlike the situation in the U.S. and in most other countries, turning a critical age cutoff (which is 18 in Germany) does not automatically imply application of the ‘adult’ law. Instead there is a smooth transition to a fuzzy design of either adult or juvenile legislation. The selection is subject to discretionary court decisions, which depend on the severity of the crime and the personal development of the defendant. This particular situation requires an innovative research design. Application of (fuzzy) regression discontinuity seems to be straightforward. However, the cutoff is hardly noticeable at the age of 18 which limits the usual advantages of RD. Thus, maximum likelihood estimations of a joint two-equation system (selection and treatment equations) have been performed in order to identify the causal effect of the ‘adult’ treatment, thereby controlling for the selectivity into ‘adult’ and ‘juvenile’ regimes.

Results suggest that inmates of the adolescent age group [18,21) convicted under adult criminal law have a lower propensity for future offending. Alcohol and drug addiction, low social capital, criminal family background and bad labor market perspectives belong to the most important factors driving expected recidivism. Inspecting the court decisions in favor of adult criminal law, robbery is most likely treated under juvenile law and intellectual and moral development of defendants seem to be best proxied by age and educational degree (‘Abitur’).

Results are in line with the hypothesis that treatment under the stricter adult criminal law has some specific deterrent effect such that inmates intend to avoid future imprisonment under such conditions. Of course, results would also be consistent with the alternative explanation that prisoner rehabilitation and corrective measures prescribed for adolescents are less

effective under juvenile law. This interpretation, however, would suggest highly inefficient measures of social integration as costs per juvenile inmate exceed those of adult criminals by more than 40% (Entorf et al. 2008).

There are some limitations of this study which could only be derestricted when large-scale individual panel data sets on recidivism were available in Germany. Unfortunately, there is no publicly available source which collects individual information on recidivism and its background factors of the German prison population. The only official source is Jehle et al. (2003) which provides aggregate statistics of recidivism realized during the short period 1994 to 1998.¹⁵ Thus, even though the number of observations is rather small and data on recidivism are based on self assessment, the present study offers a rare evaluation of factors leading judges to convict defendants under adult criminal law (instead of juvenile law), thereby allowing for testing the specific deterrence hypothesis for adolescent offenders. It is hoped that the study will encourage more research dealing with the problem of selection and recidivism, particularly in European countries.

¹⁵ Recidivism statistics do not provide further information on factors of recidivism. Even information on state-dependent recidivism rates is withheld by the German Government for political reasons: see the statement made by the Federal Office of Justice (*Bundesamt für Justiz*, Bonn) in a newspaper article (see FAZ, July 16th, 2010).

7 References

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APPENDIX

Table A1: Descriptive statistics: Total sample

Variable	Nobs	Mean	Std. Dev.	Min	Max
recidivism (expectation)	859	.2188591	.4137138	0	1
age at offense	837	31.75876	10.85896	15.3	65.5
age	859	35.13853	11.61426	16	68
male	859	.8824214	.3222961	0	1
married	859	.1734575	.3788629	0	1
poor social capital	859	.5552969	.4972224	0	1
no church	859	.2735739	.4460525	0	1
criminal family background	859	.1699651	.3758209	0	1
village	859	.2188591	.4137138	0	1
addiction (alcohol or drugs)	859	.2805588	.449534	0	1
good job market prospect	859	.2060536	.404705	0	1
schooling	847	10.34475	2.170845	2	24
length of prison sentence	838	4.449165	3.639345	.11	19
adult	859	.8719441	.3343467	0	1
murder	859	.112922	.3166818	0	1
sexual offence	859	.1233993	.3290866	0	1
assault	859	.1897555	.392336	0	1
robbery	859	.1559953	.3630623	0	1
theft	859	.233993	.4236144	0	1
fraud	859	.2153667	.411316	0	1
illicit drugs: consumption	859	.0698487	.2550403	0	1
illicit drugs: dealing	859	.1408615	.3480813	0	1
vandalism	859	.0232829	.1508884	0	1

Note: descriptive statistics under the restriction that variables show valid observations on 'recidivism'; Source: German Inmate Survey (Entorf et al, 2008)

Table A2: Descriptive statistics, age at offence = (15, 18), juvenile penal law

Variable	Nobs	Mean	Std. Dev.	Min	Max
recidivism (expectation)	55	0.2909091	0.4583678	0	1
age at offense	55	16.85832	0.8070685	15.3	17.9
age	55	18.85455	1.393346	16	23
male	55	1	0	1	1
married	55	0.0181818	0.13484	0	1
poor social capital	55	0.4727273	0.5038572	0	1
no church	55	0.3454545	0.479899	0	1
criminal family background	55	0.2181818	0.4168182	0	1
village	55	0.1272727	0.33635	0	1
addiction (alcohol or drugs)	55	0.2727273	0.4494666	0	1
job contact	55	0.2181818	0.4168182	0	1
schooling	55	9.290909	1.565516	4	12
length (of prison sentence)	55	2.907576	2.141933	0.5	9.5
(application of) criminal law	55	0	0	0	0
murder	55	0.1272727	0.33635	0	1
sexual offence	55	0.0363636	0.1889186	0	1
assault	55	0.4909091	0.504525	0	1
robbery	55	0.4363636	0.5005048	0	1
theft	55	0.5636364	0.5005048	0	1
fraud	55	0.1090909	0.3146266	0	1
illicit drugs: consumption	55	0.0363636	0.1889186	0	1
illicit drugs: dealing	55	0.1272727	0.33635	0	1
vandalism	55	0.1636364	0.373355	0	1

Note: descriptive statistics under the restriction that variables show valid observations on 'recidivism'; Source: German Inmate Survey (Entorf et al, 2008)

Table A3: Descriptive statistics, age at offence = [18, 21), young adults (juvenile or adult law)

Variable	Nobs	Mean	Std. Dev.	Min	Max
recidivism (expectation)	78	0.3205128	0.4696943	0	1
age at offense	78	19.64153	0.9005078	18	20.9
age	78	21.79487	2.188274	19	29
male	78	0.9358974	0.2465209	0	1
married	78	0.1025641	0.3053524	0	1
poor social capital	78	0.4487179	0.5005824	0	1
no church	78	0.3717949	0.4864121	0	1
criminal family background	78	0.2692308	0.4464311	0	1
village	78	0.1538462	0.3631365	0	1
addiction (alcohol or drugs)	78	0.2948718	0.4589365	0	1
good job market prospect	78	0.1923077	0.3966644	0	1
schooling	78	10.03846	2.263157	3	17
length of prison sentence	76	3.192982	2.758563	0.3	15
application of criminal law	78	0.2820513	0.4529108	0	1
murder	78	0.1282051	0.336482	0	1
sexual offence	78	0.0384615	0.1935524	0	1
assault	78	0.3974359	0.492535	0	1
robbery	78	0.2948718	0.4589365	0	1
theft	78	0.3333333	0.4744557	0	1
fraud	78	0.2179487	0.4155246	0	1
illicit drugs: consumption	78	0.0897436	0.287664	0	1
illicit drugs: dealing	78	0.1410256	0.3503008	0	1
vandalism	78	0.1282051	0.336482	0	1

Note: descriptive statistics under the restriction that variables show valid observations on 'recidivism'; Source: German Inmate Survey (Entorf et al, 2008)

Table A4: Descriptive statistics, age at offence = [21, 22], adult criminal law

Variable	Nobs	Mean	Std. Dev.	Min	Max
recidivism (expectation)	26	0.1923077	0.4019185	0	1
age at offense	26	21.54961	0.2781412	21	22
age	26	24.30769	1.783687	22	29
male	26	0.7692308	0.4296689	0	1
married	26	0.1538462	0.3679465	0	1
poor social capital	26	0.6153846	0.4961389	0	1
no church	26	0.2307692	0.4296689	0	1
criminal family background	26	0.0769231	0.2717465	0	1
village	26	0.3076923	0.4706787	0	1
addiction (alcohol or drugs)	26	0.3076923	0.4706787	0	1
good job market prospect	26	0.1538462	0.3679465	0	1
schooling	26	10.38462	2.401282	2	15
length of prison sentence	26	3.645833	3.224715	0.2	14
application of criminal law	26	1	0	1	1
murder	26	0.0769231	0.2717465	0	1
sexual offence	26	0.1153846	0.3258126	0	1
assault	26	0.1153846	0.3258126	0	1
robbery	26	0.1153846	0.3258126	0	1
theft	26	0.3461538	0.4851645	0	1
fraud	26	0.2307692	0.4296689	0	1
illicit drugs: consumption	26	0.1153846	0.3258126	0	1
illicit drugs: dealing	26	0.1538462	0.3679465	0	1
vandalism	26	0	0	0	0

Note: descriptive statistics under the restriction that variables show valid observations on 'recidivism'; Source: German Inmate Survey (Entorf et al, 2008)