

Online Appendix of “Modeling the Effects of Grade Retention in High School”

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A Proof of Proposition 1

Proof.

Identification of the unknown components in the first schooling year

First, observe that, conditional on the age at the start of high school in_i ,¹ the first high school outcome, i.e. track choice $Y_{i01}^* \equiv tr_{i1} = \mu(\mathbf{x}_i, z_i, in_i; \theta_{01}) + u_{i01}$ is free of selection and can, hence, be considered as a “measurement”. The track choice can, however, be selective in the sense that it can be related to the first end-of-year evaluation $Y_{i11}^* = ev_{i1} = \mu(\mathbf{x}_i, tr_{i1}, in_i; \theta_{11}) + u_{i11}$ through the common unobserved determinant $v_{i1}(7)$, which induces dependence between u_{i01} and u_{i11} .² By the presence of the continuous variable z_i that is excluded from all the other outcome equations (condition 1 in Proposition 1),³ one can vary track choice and, hence, u_{i01} , independently of the end-of-year evaluation in grade 7 and, therefore, independently of u_{i11} . As shown by Theorem 1 of Carneiro et al. (2003), this independent variation identifies the joint distribution of (u_{i01}, u_{i11}) non-parametrically (up to scale), and the corresponding unobserved threshold parameters of the ordered choice models $\alpha_{j,0}$ ($j \in \{1, \dots, 4\}$) and $\alpha_{k,1}$ ($k \in \{1, 2\}$).⁴ Key is that the error terms u_{i01}

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¹See further discussion of this point in Section 4.3

²By Assumption 3.1, $u_{i01} = \delta_{07}v_{i1}(7) + \epsilon_{i01}$ and $u_{i11} = \delta_{17}v_{i1}(7) + \epsilon_{i11}$, where $\delta_{17} = 1$ by normalization.

³Alternatively, in the absence of an exclusion restriction, Carneiro et al. (2003) show that it is sufficient to have two components of \mathbf{x} that vary continuously over \mathbb{R} (condition 2 in our Proposition 1).

⁴Theorem 1 actually assume that the discrete measurements are binary valued and not ordered choices. However, on p. 376 the authors state that the extension to censored random variables, i.e. ordered choices, is straightforward.

and u_{i11} (and the error terms of all other outcomes)⁵ do not depend on track choice. This means that there cannot be any higher order dependence that is not captured by the conditional mean, and that the factor loading δ_{11} on the unobserved determinant $v_{i1}(7)$ cannot depend on tr_i , i.e. this excludes *essential heterogeneity* in the effect of track choice.

Based on this result, identification of the unknown components of the error terms u_{i01} and u_{i11} is shown in three steps. First, since by the aforementioned argument the joint distribution of the error terms of the first two outcomes is identified, we can form the following ratio of cross moments (where $\delta_{17} = 1$ by normalization):⁶

$$\frac{E[u_{i01}^3 u_{i11}]}{E[u_{i01} u_{i11}^3]} = \frac{\delta_{07}^3 E[v_{i1}(7)^4]}{\delta_{07} E[v_{i1}(7)^4]} = \delta_{07}^2 \quad (\text{A-1})$$

using that $u_{i01} = \delta_{07} v_{i1}(7) + \epsilon_{i01}$, $u_{i11} = \delta_{17} v_{i1}(7) + \epsilon_{i11}$, $\delta_{07} \neq 0$, Assumptions 1.2 and 3.2, and the existence of the fourth order moment of the distribution of $v_{i1}(7)$. This identifies δ_{07} apart from its sign. The sign of δ_{07} corresponds to the sign of $E[u_{i01} u_{i11}] = \delta_{07} E[v_{i1}(7)^2]$, since $E[v_{i1}(7)^2] > 0$ and finite.

Second, from the the higher order cross moments of the residuals of the two outcome equations, we can recover all moments of the the unobserved determinants $v_{i1}(7)$: $\forall k > 0$: $E[(\delta_{07} v_{i1}(7) + \epsilon_{i01})^k (\delta_{17} v_{i1}(7) + \epsilon_{i11})] = \delta_{07}^k E[v_{i1}(7)^{k+1}]$. Since δ_{07} is already identified and since a distribution of which all moments are finite (Assumption (iv) in Proposition 1) can be completely characterized by these moments (Billingsley, 1995), this non-parametrically identifies the distribution of $v_{i1}(7)$.

Finally, we can form all the higher order moments of the error terms of each of the two outcome equations, $\forall k > 1$: $E[(\delta_{07} v_{i1}(7) + \epsilon_{i01})^k] = \delta_{07}^k E[v_{i1}(7)^k] + E[\epsilon_{i01}^k]$ and $E[(\delta_{11} v_{i1}(7) + \epsilon_{i11})^k] = E[v_{i1}(7)^k] + E[\epsilon_{i11}^k]$. Since the first terms of the sum on the right hand-side are already identified, the second terms can be recovered. This enables non-parametric identification of the distribution of ϵ_{i01} and ϵ_{i11} .

In first period the next outcome is the decision to repeat the grade or not ($Y_{i31} \equiv re_{i1}$) for stu-

Note also that FNT can prove identification of the joint distribution of these error terms without exclusion restriction and without the presence of continuous explanatory variables, because they have continuous outcome variables, i.e. test scores. In the case of continuous outcomes the joint distribution of the error terms can be identified by constructing all (cross) moments of the residuals in the outcome equations. In the case of discrete outcomes, these residuals are “latent”, so that their cross moments cannot be directly formed and, hence, stronger identifying assumptions are required.

⁵Further on in the proof we repeatedly use Theorem 1 of Carneiro et al. (2003) to prove identification of the joint distribution of u_{i01} and the error terms u_{ict} of each of the other outcomes for $c \in \{1, \dots, 4\}$ and $t \in \{1, \dots, T_i\}$.

⁶Instead of forming the ratio of fourth order moments FNT consider the ratio of third order moments. This identification argument works only if $E[v_{i1}(7)^3] \neq 0$ (and, hence only for asymmetric distributions), because this third moment appears then in the denominator of the ratio. In their Appendix B FNT relax this asymmetry assumption in the case of having measurements of more test scores per student, also using fourth moments. Our identification strategy is inspired by a combination of the arguments mentioned in their main text and in their Appendix B.

dents for whom $Y_{i11} \equiv ev_{i1} = B$. This is because nobody drops out in the seventh grade ($out_{i1} \equiv Y_{i21} = 0$).⁷ Using again Theorem 1 of Carneiro et al. (2003) θ_{31} , the associated threshold parameters, and the joint distribution of (u_{i01}, u_{i31}) are identified. The latter allows us to form the following cross moment: $E[u_{i01}u_{i31}] = E[(\delta_{07}v_{i1}(7) + \epsilon_{i01})(\delta_{37}v_{i1}(7) + \epsilon_{i31})] = \delta_{07}\delta_{37}E(v_{i1}(7)^2)$. δ_{37} is identified, because $\delta_{07} \neq 0$ and $E(v_{i1}(7)^2) \neq 0$ already are. As before, the higher order moments of the third outcome equation then identifies the distribution of ϵ_{i31} . Following the same argument θ_{41} , the associated threshold parameters and the distribution of ϵ_{i41} are identified as well.

Identification of the unknown components beyond the first schooling year

As from period 2 some pupils may have been retained. This means that u_{ic2} depends on re_{i1} through $\delta_{c8}(re_{i1})$ (Assumption 3.1). A consequence is that to identify $\delta_{c8}(re_{i1})$ we have to condition on two sub-populations: the population that has been retained in the previous year ($re_{i1} = 1$) and the one that has not been retained ($re_{i1} = 0$). This is possible (cf. next paragraph) because the selection into retention occurs through dependence on observables and *past* unobservables that have already been identified, while the distribution of the new unobserved persistent shock in grade 8 $v_{i1}^*(8)$ can be identified from the cross moments between the unobservables of individuals who are retained in grade 8, but not in grade 7.

First, consider the error of outcome c for an individual i who is retained in grade 7 ($re_{i1} = 1$). From Assumptions 3.1 and 3.2 we obtain: $u_{ic2} = \delta_{c7}(1)v_{i1}(7) + \epsilon_{ic2}$. We can then apply, as in period 1, Theorem 1 of Carneiro et al. (2003) to prove that the joint conditional distribution $(u_{i01}, u_{ic2})|re_{i1} = 1, \theta_{c2}$ for $re_{i1} = 1$ and the threshold parameters of the corresponding ordered choice are non-parametrically identified. Hence, we can form the following conditional cross moment: $E(u_{i01}, u_{ic2}|re_{i1} = 1) = \delta_{07}\delta_{c7}(1)E(v_{i1}(7)^2)$. Since δ_{07} and the distribution of $v_{i1}(7)$ is already identified, this cross moment identifies $\delta_{c7}(1)$.

Second, consider the error of outcome $c = 1$ for an individual who is not retained in grade 7 ($re_{i1} = 0$): $u_{i12} = \delta_{18}(0)(\delta_7^*(8)v_{i1}(7) + v_{i1}^*(8))$. Noticing that $\delta_{18}(0) = 1$ by normalization (Assumption 3.1), we can follow a similar argument as in the previous paragraph to show that $\delta_7^*(8), \theta_{12}$ for $re_{i1} = 0$ and the threshold parameters of the corresponding ordered choice. Once $\delta_7^*(8)$ is identified, we can follow a same strategy for outcomes $c \neq 1$ to identify $\delta_{c8}(0), \theta_{c2}$ for $re_{i1} = 0$ and the corresponding threshold parameters.

In order to identify the distribution of the new persistent shock $v_{i1}^*(8)$, we consider the error of outcome c for an individual who is retained in grade 8, but not in grade 7: $u_{ic3} = \delta_{c8}(0, 1)(\delta_7^*(8)v_{i1}(7) + v_{i1}^*(8))$. Following a similar argument as in the previous paragraphs we can first form the following cross moments: $E(u_{i01}u_{ic3}|re_{i1} = 0, re_{i2} = 1) = \delta_{07}\delta_{c8}(0, 1)\delta_7^*(8)E(v_{i1}(7)^2)$. Since

⁷In our data the first grade repetition occurs only in grade 8, so that this issue starts only as of period 3. We ignore this here, to demonstrate that identification does not hinge on this particularity. In this case the next outcome in the first period would rather be $dow_{i1} \equiv Y_{i41}$.

δ_{07} , $\delta_7^*(8)$ and the distribution of v_{i1} are already identified these identify $\delta_{c8}(0\ 1)$. Next we form the cross moments of the error terms of in grade 8 of individuals who have been retained in that grade (but not in grade 7): $E(u_{ic2}^k u_{ic3} | re_{i1} = 0, re_{i2} = 1) = \delta_{c8}(0\ 1) \delta_7^*(8)^{k+1} E(v_{i1}(7)^{k+1}) + \delta_{c8}(0\ 1) E(v_{i1}^*(8)^{k+1})$ for $k > 0$. Since only $E(v_{i1}^*(8)^{k+1})$ is the only unknown in this expression, these cross moments identify the distribution of $v_{i1}^*(8)$.

We can proceed in a similar way sequentially over time periods, outcomes, and retention histories until we arrive at the end of the observation period to identify to full joint distribution of grade-varying unobserved heterogeneity \mathbf{v}_i , all θ_{ct} and associated threshold parameters. ■

B Partial Observability of Track Choices at the Start of High School

In Subsection 4.5 we explained that we do not observe the chosen track at the beginning of high school, i.e. tr_{i1} . Mroz et al. (2016) solve this partial observability by considering the marginal likelihood function instead of the conditional one, where the unobserved information is integrated out of the likelihood. Here we follow a similar approach by summing the likelihood over the possible initial track choices at the start of high school. As in Mroz et al. (2016), we take prior information into account to restrict the potential number of initial track choices tr_{i1} over which we sum the likelihood. The following prior information is considered: (i) the initial track choice of pupils starting in the vocational track, i.e. for whom $tr_{i1} = 1$, is observed; (ii) the track choice is known in all grades beyond grade 7; (iii) no student is retained in grade 7 ($re_{i1} = 0$); (iv) as all students are younger than 18 years old in grade 7, no student drops out high school in this grade ($out_{i1} = 0$); (v) students can only downgrade ($0 \leq dow_{i1}$) and if they do, they do not downgrade more than two tracks in a single year ($dow_{i1} \leq 2$). In this appendix, we show how the marginal likelihood function that accommodates for the partial observability of tr_{i1} can be adjusted to take this prior information into account.

First, to focus on the main issues, we simplify the notation. We ignore in the joint distribution function as expressed by Equation (1) in Section 4.1 the subscripts and the conditioning on in_i and the observed and unobserved covariates:

$$D(tr_1, \mathbf{Y}) = D(tr_1)D(ev_1, dow_1 | tr_1)D(Y_2 \dots Y_T | ev_1, dow_1, tr_1) \quad (\text{A-2})$$

where $D(\cdot)$ and $D(\cdot, \cdot | \cdot)$ respectively denote the marginal and (joint) conditional distributions of their arguments and where we recall that $Y_1 = [ev_1\ 0\ 0\ dow_1] \equiv [ev_1\ dow_1]$, because, by (iv), nobody drops out in grade 7 ($out_1 = 0$), and, by (iii), nobody is retained in grade 7 ($re_1 = 0$). $D(Y_2 \dots Y_T | ev_1, dow_1, tr_1)$ in (A-2) (and also $D(dow_1 + tr_2, dow_1 | ev_1, tr_2)$ in (A-4) below) are conditional on $out_1 = 0$ and $re_1 = 0$. However, in order to avoid burdensome notation we leave

this conditioning implicit.

In order to take partial observability of the initial track choice into account, we should sum the joint distribution in Equation (A-2) over tr_1 for all pupils who are not in the vocational track, i.e. for whom $tr_1 > 1$. However, given the available prior information, the sum should not be over all four unknown tracks ($2 \leq tr_1 \leq 5$). Recall that by prior information (ii) the track choice in grade 8 (tr_2) is known. Together with the fact (v) that pupils can only downgrade, and if they downgrade, they can downgrade at most two tracks ($0 \leq dow_{i1} \leq 2$), the initial track choice tr_1 is restricted, depending on the track choice tr_2 observed in grade 8.

To see more clearly how the prior information (ii) and (v) restricts the number of tracks over which the joint distribution (A-2) is summed, note first that we do not observe dow_t in grade t directly, but we infer it from the tracks in which pupils are observed in each year beyond grade 7:⁸

$$dow_t = tr_t - tr_{t+1} \quad (\text{A-3})$$

This equation establishes a one-to-one relationship between tr_1 and dow_1 for any given value of tr_2 : $tr_1 = dow_1 + tr_2$. This means that if we condition (A-2) on the known value of tr_2 ,⁹ summing this equation over the unknown tr_1 is equivalent to summing it over the unknown dow_1 . The advantage of summing it over dow_1 is that we can easily impose the prior information that both $0 \leq dow_{i1} \leq 2$ and $(2 \leq tr_1 = dow_1 + tr_2 \leq 5 \Leftrightarrow 2 - tr_2 \leq dow_1 = tr_1 - tr_2 \leq 5 - tr_2)$ by setting $\max\{0, 2 - tr_2\} \leq dow_1 \leq \min\{2, 5 - tr_2\}$. In case $tr_2 = 1$, $dow_1 > 0$, because $tr_1 \neq 1$, as we observe the track choice for individuals in VHS at the start, i.e. for $tr_1 = 1$.

In order to take the partial observability for $tr_1 > 1$ into account, we therefore consider Equation (A-2) given tr_2 , replace tr_1 by $dow_1 + tr_2$ and sum it, instead of over tr_1 , over all possible downgrading decisions dow_1 , taking the prior information into account:

$$\begin{aligned} & \sum_{dow_1 = \max\{0, 2 - tr_2\}}^{\min\{2, 5 - tr_2\}} D(dow_1 + tr_2, \mathbf{Y}) = \\ & \sum_{dow_1 = \max\{0, 2 - tr_2\}}^{\min\{2, 5 - tr_2\}} D(dow_1 + tr_2) D(ev_1, dow_1 | dow_1 + tr_2) \\ & \quad \times D(Y_2 \dots Y_T | dow_1 + tr_2, ev_1, dow_1). \end{aligned} \quad (\text{A-4})$$

The sample log-likelihood function in Equation (4) in the main text is modified along these lines.

⁸As no student is retained in grade 7, we observe all track choices for $t > 1$ and $g > 7$.

⁹Note that if both tr_1 and dow_1 are known, (A-3) implies that tr_2 is irrelevant, because it does not add any new information. By contrast, if neither tr_1 nor dow_1 are known, as in the case of partial observability, tr_2 matters, because it adds in new information. That is why it appears when summing over dow_1 (which is equivalent to summing over tr_1) in Equation (A-4) below, while it is absent in (A-2).

C The Empirical Specification of the Educational Choices

As mentioned in Subsection 4.4 we assume that all educational choices can be specified as (ordered) logits. As discussed in Subsection 4.3, this is not strictly required for identification. In the following subsections we first describe in detail for each schooling outcome the model specification choices. In the final subsection we discuss the specification of the joint unobserved heterogeneity distribution $G(\mathbf{v}_{i1}; \rho)$.

C.1 The Track Choice at the Start of High School

The track choice takes value on $\{VHS, THS^-, THS^+, GHS^-, GHS^+\}$, which we relabel for notational convenience and to underline their hierarchical ordering by $\{1, 2, 3, 4, 5\}$. The probability density function of the track choice is an ordered logit determined by a linear index in the strictly exogenous observed explanatory variables $\mathbf{z}_i \equiv [z_i \ \mathbf{x}_i]$, including the ‘‘instrument’’ z_i , i.e. the day of birth, which is excluded from all other schooling outcome equations, the unobservable determinants of track choice $\mathbf{v}_{i,tr} \equiv \mathbf{v}_{i0} \equiv v_{i0}(7) = \delta_{tr}v_{i1}(7)$, where $\delta_{tr} \equiv \delta_0$ is the loading factor of the unobserved heterogeneity, and the initial delay in_i :

$$\begin{aligned}
 \Pr(tr_{i1} = 1 | \mathbf{z}_i, \mathbf{v}_{i,tr}, in_i) &= \Lambda(\alpha_{1,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}), \\
 \Pr(tr_{i1} = 2 | \mathbf{z}_i, \mathbf{v}_{i,tr}, in_i) &= \Lambda(\alpha_{2,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}) - \Lambda(\alpha_{1,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}), \\
 \Pr(tr_{i1} = 3 | \mathbf{z}_i, \mathbf{v}_{i,tr}, in_i) &= \Lambda(\alpha_{3,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}) - \Lambda(\alpha_{2,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}), \\
 \Pr(tr_{i1} = 4 | \mathbf{z}_i, \mathbf{v}_{i,tr}, in_i) &= \Lambda(\alpha_{4,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}) - \Lambda(\alpha_{3,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}), \\
 \Pr(tr_{i1} = 5 | \mathbf{z}_i, \mathbf{v}_{i,tr}, in_i) &= 1 - \Lambda(\alpha_{4,tr} - \mathbf{v}_{i,tr} - \mathbf{z}'_i \boldsymbol{\beta}_{tr} - in_i \gamma_{tr}), \tag{A-5}
 \end{aligned}$$

where δ_{tr} , $\boldsymbol{\beta}_{tr}$ and γ_{tr} are parameters to be estimated, the coefficients $\alpha_{1,tr} < \alpha_{2,tr} < \alpha_{3,tr} < \alpha_{4,tr}$ are the ordered threshold parameters and $\Lambda(\cdot)$ denotes the logistic distribution.

C.2 The End-of-Year Evaluation

At the end of each academic year, teachers jointly evaluate in a staff meeting the global academic performance of the pupils in the past year. As mentioned in Section 2, students obtain one of the following three scores: A, B or C. An A allows students to be promoted to the next grade. Students getting a C must repeat the grade. Students with a B are imposed to downgrade the track, unless they accept to repeat the grade, in which case they can freely choose to downgrade or not. Because of the natural ordering of these scores, the staff’s evaluation choices are modeled as an ordered logit, conditional on both the strictly exogenous observed and unobserved explanatory variables

and the past educational choices of pupils and teachers:

$$\begin{aligned}
\Pr(ev_{it} = C | \mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) &= \Lambda[\alpha_{1,ev} - \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})], \\
\Pr(ev_{it} = B | \mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) &= \Lambda[\alpha_{2,ev} - \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})] \\
&\quad - \Lambda[\alpha_{1,ev} - \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})], \\
\Pr(ev_{it} = A | \mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) &= 1 - \Lambda[\alpha_{2,ev} - \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})],
\end{aligned} \tag{A-6}$$

where $\phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})$ is a linear index in its arguments, capturing the impact of observed and unobserved determinants and past educational choices of pupils and teachers.

We consider two different specifications of the linear index $\phi_{ev}(\cdot)$. The first specification ignores *essential* heterogeneity in the effect of the past grade repetition on the evaluation outcome, while the second one explicitly allows for it. In the first specification the linear index takes the following form:

$$\begin{aligned}
\phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) &= \mathbf{x}'_i \boldsymbol{\beta}_{ev} + \boldsymbol{\iota}'_{it} \mathbf{v}_{i,ev} + in_i \gamma_{ev} + \mathbf{Idow}'_{it-1} \boldsymbol{\pi}_{ev} + \mathbf{Itr}'_{it} \boldsymbol{\eta}_{ev} \\
&\quad + \mathbf{Igr}'_{it} \boldsymbol{\delta}_{ev} + re_{it-1} \kappa_{ev} + pre_{it-1} \psi_{ev} \\
&\equiv \mathbf{w}'_{it} \boldsymbol{\xi}_{ev} + re_{it-1} \kappa_{ev} + pre_{it-1} \psi_{ev},
\end{aligned} \tag{A-7}$$

where

- $\mathbf{v}_{i,ev} \equiv \mathbf{v}_{i1}$ is the vector of unobserved heterogeneity affecting the evaluation;
- $\boldsymbol{\iota}'_{it} = [1 \ \mathbf{Igr}'_{it}]$ is a 1×7 indicator vector selecting the unobserved heterogeneity component associated to the grade in which individual i is in period t (taking grade 7 as the reference);
- $\mathbf{Igr}_{it} = \left[\mathbf{1}_{\{8\}} \left(6 + t - \mathbf{1}_{\{\forall t:t>1\}}(t) \sum_{s=1}^{t-1} re_{is} \right) \dots \mathbf{1}_{\{13\}} \left(6 + t - \mathbf{1}_{\{\forall t:t>1\}}(t) \sum_{s=1}^{t-1} re_{is} \right) \right]'$ is a column vector of six indicators of the grade at the beginning of the t -th year in high school and where grade 7 is the reference grade;
- $\mathbf{Idow}_{it-1} = [\mathbf{1}_1(dow_{it-1}) \ \mathbf{1}_2(dow_{it-1})]'$ is a column vector of two indicators that determine whether the student chooses to downgrade one or two tracks at the end of the previous academic year (the reference student does not change track): $\mathbf{1}_A(x)$ defines the indicator function that is equal to one if $x \in A$ and zero otherwise; $dow_{it} \in \{0, 1, 2\}$ indicates the number of tracks that individual i chooses to downgrade at the end of year t in high school;
- $\mathbf{Itr}_{it} = \left[\mathbf{1}_{\{2\}} \left(tr_{i1} - \mathbf{1}_{\{\forall t:t>1\}}(t) \sum_{s=1}^{t-1} dow_{is} \right) \dots \mathbf{1}_{\{5\}} \left(tr_{i1} - \mathbf{1}_{\{\forall t:t>1\}}(t) \sum_{s=1}^{t-1} dow_{is} \right) \right]'$ is a column vector of four indicators of the track chosen at the beginning of the t -th year in high school and where VHS ($tr_{i1} = 1$) is the reference track;

- re_{it-1} is an indicator variable equal to one if individual i was retained at the end of the previous academic year $t - 1$ (repeating the grade therefore in the current year t) and zero otherwise;
- $pre_{it-1} = \mathbf{1}_{\mathbb{N}_0} \left(\mathbf{1}_{\{\forall t:t>2\}}(t) \sum_{s=1}^{t-2} re_{is} \right)$ is an indicator equal to one if the student has ever repeated a grade in high school in years prior to the $(t - 1)^{th}$ year.
- $\mathbf{w}'_{it} \equiv [\mathbf{x}'_i \ \iota'_{it} \ in_i \ \mathbf{Idow}'_{it-1} \ \mathbf{Itr}'_{it} \ \mathbf{Igr}'_{it}]$ and $\boldsymbol{\xi}'_{ev} = [\boldsymbol{\beta}'_{ev} \ \mathbf{v}'_{i,ev} \ \gamma_{ev} \ \boldsymbol{\pi}'_{ev} \ \boldsymbol{\eta}'_{ev} \ \boldsymbol{\delta}'_{ev}]$

where $\boldsymbol{\xi}'_{ev}$, κ_{ev} and ψ_{ev} are parameters to be estimated.¹⁰ Hence, we allow that past high school choices (\mathfrak{S}_{it-1}) affect the evaluations in a flexible way. The coefficient κ_{ev} is the transitory effect of grade repetition on the subsequent academic performance, while ψ_{ev} is the permanent effect. $\boldsymbol{\delta}_{ev}$ and $\boldsymbol{\eta}_{ev}$ capture that students' ability to get good evaluations depends on the current grade and track, respectively. Finally, $\boldsymbol{\pi}_{ev}$ is the (transitory) effect of having downgraded a track on the academic achievement in the subsequent year.

In the second specification of the linear index $\phi_{ev}(\cdot)$ we allow the short- and long-run effects of grade repetition to be heterogeneous in observed and unobserved abilities. To maintain a tractable model, we simplify by interacting re_{it-1} and pre_{it-1} with the linear index defined in Equation (A-7). Since this linear index is also a function of unobservables, this allows for *essential* heterogeneity in the treatment effect of grade repetition. The specification of the linear index is then given by the following expression:

$$\begin{aligned} \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) &= \mathbf{w}'_{it} \boldsymbol{\xi}_{ev} (1 + re_{it-1} \kappa_{ev}^0 + pre_{it-1} \psi_{ev}^0) \\ &\quad + re_{it-1} \kappa_{ev} (1 + pre_{it-1} \psi_{ev}^0) + pre_{it-1} \psi_{ev} (1 + re_{it-1} \kappa_{ev}^0), \end{aligned} \quad (\text{A-8})$$

where κ_{ev}^0 and ψ_{ev}^0 are parameters to be estimated. If κ_{ev}^0 and ψ_{ev}^0 are jointly equal to 0, then we go back to the first specification of the linear index in Equation (A-7).

In the last two grades or if a student is in the VHS track, the evaluation is dichotomous, either A or C. In these cases, the ordered logit model described in Equation (A-6) collapses to a logit model with:

$$\Pr(ev_{it} = A | \mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) = 1 - \Lambda[\alpha_{2,ev} - \phi_{ev}(\mathbf{x}_i, \mathbf{v}_{i,ev}, in_i, tr_{i1}, \mathfrak{S}_{it-1})], \quad (\text{A-9})$$

and the probability of getting a C is its complement.

¹⁰ $\mathbf{v}'_{i,ev}$ is a function of parameters once it is replaced by the values of the corresponding points of support.

C.3 The School Drop-Out

In Belgium, compulsory education ends on 30 June of the year in which a student turns 18. From that date onwards, students are at risk of high school drop-out without diploma. School drop-out is an interesting long-run outcome of grade repetition that, as mentioned in the Introduction, other authors have considered as well. We model it as a binary choice in the following way for pupils at risk ($s_{it} = 1$):¹¹

$$\Pr(out_{it} = 1 | \mathbf{x}_i, \mathbf{v}_{i,out}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}) = \Lambda [\alpha_{out} + \phi_{out}(\mathbf{x}_i, \mathbf{v}_{i,out}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it})], \quad (\text{A-10})$$

where $\mathbf{v}_{i,out} \equiv \mathbf{v}_{i2} = \delta_{out} \mathbf{v}_{i1}$, $\delta_{out} \equiv \delta_2$ is the loading factor of the unobserved heterogeneity distribution, and where similar to Equation (A-8),

$$\begin{aligned} \phi_{out}(\mathbf{x}_i, \mathbf{v}_{i,out}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}) &= (\mathbf{w}'_{it} \boldsymbol{\xi}_{out} + \mathbf{I} \mathbf{ev}'_{it} \boldsymbol{\omega}_{out}) (1 + re_{it-1} \kappa_{out}^0 + pre_{it-1} \psi_{out}^0) \\ &+ re_{it-1} \kappa_{out} (1 + pre_{it-1} \psi_{out}^0) + pre_{it-1} \psi_{out} (1 + re_{it-1} \kappa_{out}^0), \end{aligned} \quad (\text{A-11})$$

where $\boldsymbol{\xi}_{out} \equiv [\beta'_{out} \mathbf{v}'_{i,out} \gamma_{out} \boldsymbol{\pi}'_{out} \boldsymbol{\eta}'_{out} \boldsymbol{\delta}'_{out}]$, $\boldsymbol{\omega}_{out}$, κ_{out} , ψ_{out} , ψ_{out}^0 and κ_{out}^0 are parameters to be estimated and $\mathbf{I} \mathbf{ev}_{it} = [\mathbf{1}_{\{A\}}(ev_{it}) \mathbf{1}_{\{B\}}(ev_{it})]$.¹² Compared to Equation (A-8), ϕ_{out} has the extra argument, $\mathbf{I} \mathbf{ev}_{it}$, i.e. the end-of-year evaluation. By the sequential ordering assumed in Assumption 2, $\mathbf{I} \mathbf{ev}_{it}$ is predetermined with respect to the drop-out choice, so that it can be conditioned upon.

C.4 The Choice of Repeating the Grade in Case of a B Evaluation

Students getting a B can choose either to repeat the grade or to downgrade the track. The choice is binary and, conditional on getting a B and on not dropping-out, the probability of repeating the grade is specified as follows:

$$\begin{aligned} \Pr(re_{it} = 1 | \mathbf{x}_i, \mathbf{v}_{i,re}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it} = B, out_{it} = 0) &= \\ &\Lambda [\alpha_{re} + \phi_{re}(\mathbf{x}_i, \mathbf{v}_{i,re}, in_i, tr_{i1}, \mathfrak{S}_{it-1})]. \end{aligned} \quad (\text{A-12})$$

¹¹Very few students (71, 1.7% of the sample) drop-out of school before the end of the academic year. In order to simplify the model and the timing of events, in these cases we bring forward the drop-out date at the end of the previous academic year, disregarding information on retention and track downgrade of the uncompleted academic year.

¹²Because of the limited number of students at risk of a drop-out decision, estimation was only possible if we grouped students with a B and a C into one category, so that for the drop-out decision the indicator $\mathbf{1}_{\{B\}}(ev_{it})$ was excluded. For similar reasons a coarser grouping was also imposed on $\mathbf{I} \mathbf{dow}_{it-1}$, $\mathbf{I} \mathbf{tr}_{it}$ and $\mathbf{I} \mathbf{gr}_{it}$. See the results in Section C for more details.

where $\mathbf{v}_{i,re} \equiv \mathbf{v}_{i3} = \delta_{re} \mathbf{v}_{i1}$ and $\delta_{re} \equiv \delta_3$ is the loading factor of the unobserved heterogeneity distribution. Because we do not have enough B observations to empirically identify heterogeneous effects of past grade repetition on the current decision to repeat the grade, the function $\phi_{re}(\mathbf{x}_i, \mathbf{v}_{i,re}, in_i, tr_{i1}, \mathfrak{S}_{it-1})$ is parameterized as in Equation (A-7), excluding thereby the possibility of heterogeneous retention effects across individuals:

$$\phi_{re}(\mathbf{x}_i, \mathbf{v}_{i,re}, in_i, tr_{i1}, \mathfrak{S}_{it-1}) = \mathbf{w}'_{it} \boldsymbol{\xi}_{re} + re_{it-1} \kappa_{re} + pre_{it-1} \psi_{re}. \quad (\text{A-13})$$

where $\boldsymbol{\xi}_{re} \equiv [\beta'_{re} \mathbf{v}'_{i,re} \gamma_{re} \boldsymbol{\pi}'_{re} \boldsymbol{\eta}'_{re} \delta'_{re}]$, κ_{re} and ψ_{re} are parameters to be estimated. Note that the choice of repeating the grade must be considered neither for students in VHS nor for those in the last grade, because these students may never obtain a B (see Section 2).

C.5 The Track Downgrade

In Belgium, at the beginning of high school, students can choose among different tracks characterized by different curricula. This tracking system is aimed at grouping students with similar abilities and preferences. Choosing the right track is important as it will determine future work and education opportunities. In Belgium track choice matters particularly, because tracks are hierarchically ordered and students can only move down the hierarchy. The Belgian system of tracking is therefore often referred to as a ‘cascade’ system.

We model track transitions by defining a categorical ordered dependent variable for track downgrade. As already mentioned in Section C.2, the variable of interest is denoted as $dow_{it} \in \{0, 1, 2\}$. The values reflect the three possible choices: no downgrade, one-step downgrade and two-step downgrade. Students in the VHS track are already at the bottom of the cascade and cannot downgrade further. Consequently, we model track downgrade only for GHS/THS students. The probability of a track downgrade for GHS and THS⁺ students is specified as:

$$\begin{aligned} \Pr(dow_{it} = 0 | \mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}, out_{it} = 0) &= \\ &\Lambda[\alpha_{1,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})], \\ \Pr(dow_{it} = 1 | \mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}, out_{it} = 0) &= \\ &\Lambda[\alpha_{2,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})] \\ &- \Lambda[\alpha_{1,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})], \\ \Pr(dow_{it} = 2 | \mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}, out_{it} = 0) &= \\ &1 - \Lambda[\alpha_{2,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})], \end{aligned} \quad (\text{A-14})$$

where $\mathbf{v}_{i,dow} \equiv \mathbf{v}_{i4} = \delta_{dow} \mathbf{v}_{i1}$ and $\delta_{dow} \equiv \delta_4$ is the loading factor of the unobserved heterogeneity

distribution.

The function $\phi_{dow}(\cdot)$ is a linear index similar to the one specified in Equation (A-8):

$$\begin{aligned}\phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}) &= (\mathbf{w}'_{it}\boldsymbol{\xi}_{dow} + \mathbf{I}e\mathbf{v}'_{it}\boldsymbol{\omega}_{dow} + re_{it}\tau_{dow} + re_{it}\mathbf{I}e\mathbf{v}'_{it}\boldsymbol{\zeta}_{dow}) \\ &\times (1 + re_{it-1}\kappa_{dow}^0 + pre_{it-1}\psi_{dow}^0) \\ &+ re_{it-1}\kappa_{dow}(1 + pre_{it-1}\psi_{dow}^0) \\ &+ pre_{it-1}\psi_{dow}(1 + re_{it-1}\kappa_{dow}^0).\end{aligned}\quad (\text{A-15})$$

where $\boldsymbol{\xi}_{dow} \equiv [\boldsymbol{\beta}'_{dow} \mathbf{v}'_{i,dow} \gamma_{dow} \boldsymbol{\pi}'_{dow} \boldsymbol{\eta}'_{dow} \boldsymbol{\delta}'_{dow}]$, $\boldsymbol{\omega}_{dow}$, τ_{dow} , $\boldsymbol{\zeta}_{dow}$, ψ_{dow} , κ_{dow} , ψ_{dow}^0 and κ_{dow}^0 are parameters to be estimated, and $re_{it} = 1$ for students with a B evaluation who decided to repeat the grade and $re_{it} = 0$ otherwise. As a consequence of Assumption 1, re_{it} and ev_{it} are predetermined and, hence, can be conditioned upon. We also allow for interactions between the latter two variables.

For particular groups of students the choice set is reduced. First, students in THS^- cannot make a two-step downgrade: $dow_{it} \in \{0, 1\}$. Hence, for these students the ordered logit reduces to a standard logit:

$$\begin{aligned}\Pr(dow_{it} = 0 | \mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}, out_{it} = 0) &= \\ &\Lambda[\alpha_{1,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})],\end{aligned}\quad (\text{A-16})$$

and the probability of making a one-step downgrade is equal to its complement.

Second, students with a B choosing to promote to the next grade are forced to downgrade, so that $dow_{it} \in \{1, 2\}$.¹³ Also in this case the ordered logit simplifies to:

$$\begin{aligned}\Pr(dow_{it} = 2 | \mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it}, out_{it} = 0) &= \\ &1 - \Lambda[\alpha_{2,dow} - \phi_{dow}(\mathbf{x}_i, \mathbf{v}_{i,dow}, in_i, tr_{i1}, \mathfrak{S}_{it-1}, ev_{it}, re_{it})],\end{aligned}\quad (\text{A-17})$$

and the probability of making a one-step downgrade is equal to its complement.

C.6 The Specification of the Unobserved Heterogeneity Distribution

In Section 4 we imposed a one factor specification on the distribution of unobserved heterogeneity as to be able to identify essential and grade-varying heterogeneity. The unobserved grade-varying factor \mathbf{v}_{i1} have therefore seven dimensions, one for each grade ($g = 7, \dots, 13$). Since only VHS pupils have to attend grade 13 to get the diploma (in all the other tracks the diploma is obtained

¹³Students in THS^- who are promoted to next grade are forced to downgrade and, hence, the downgrading choice is not modeled for these students.

at end of grade 12), we decided to constrain $v_{i1}(12)$ to be equal to $v_{i1}(13)$. The corresponding distribution $G(\mathbf{v}_{i1}; \boldsymbol{\rho})$ is assumed to be discrete, with a finite and, a priori, unknown number M of points of support, which each are vectors of real numbers of dimension 7×1 . This distribution assigns with a probability $p^m \equiv \Pr(\mathbf{v}_{i1} = \mathbf{v}_1^m)$ (with $\sum_{j=1}^M p^j = 1$) the vector of unobserved heterogeneity terms over all grades \mathbf{v}'_{i1} to the vector value of the m^{th} point of support:

$$\mathbf{v}'_{i1} = \mathbf{v}_1^m \equiv [v_1^m(7) \ v_1^m(8) \ v_1^m(9) \ v_1^m(10) \ v_1^m(11) \ v_1^m(12) \ v_1^m(12)]$$

where we have set $v_1^m(13) = v_1^m(12)$ and $m \in \{1, \dots, M\}$. By allowing all numbers $v_1^m(g)$ for $g \in \{7, \dots, 12\}$ to take on unrestricted values in the set of real numbers, we permit an arbitrary dependence structure of the unobserved heterogeneity between grades g .

We follow the recommendation of [Gaure et al. \(2007\)](#) by determining the number M of points of support of this distribution on the basis of the Akaike Information Criterion (AIC). The probabilities associated to the points of support sum to one and are specified as logistic transforms:

$$p^m \equiv \Pr(\mathbf{v}_{i1} = \mathbf{v}_1^m) = \frac{\exp(\rho^m)}{\sum_{h=1}^M \exp(\rho^h)} \quad \text{with } m = 1, \dots, M \quad \text{and } \rho_M = 0. \quad (\text{A-18})$$

The sample log-likelihood function in Equation (4) can be rewritten by replacing the integral by the following summation over all M points of support:

$$\ell(\boldsymbol{\theta}, \boldsymbol{\rho}) = \sum_{i=1}^N \ln \left[\sum_{m=1}^M p^m \mathcal{L}_{im}(\boldsymbol{\theta}, \boldsymbol{\rho}) \right], \quad (\text{A-19})$$

where $\mathcal{L}_{im}(\boldsymbol{\theta}, \boldsymbol{\rho})$ is the individual contribution to the likelihood function if the individual is of type m .

D Estimation Results of the Benchmark Model with and without unobserved heterogeneity

D.1 Estimated Probability Masses of the Discrete Unobserved Heterogeneity Distribution

In Table [A-1](#) we report the estimated probability masses of each point of support and other statistics of the estimated models. The number of points of support are chosen so that to minimize the Akaike Information Criterion (AIC). The resulting number of support points is $M = 3$ for both the specification controlling for grade-constant unobserved heterogeneity and the one controlling for grade-varying unobserved heterogeneity. The preferred model according to the AIC is the one

with grade-varying unobserved heterogeneity. This is the reason why in our article we consider it as the benchmark model. The location of the support points and the loading factors of each equation are reported in the next Tables.

Table A-1: Estimated Probability Masses of the Discrete Unobserved Heterogeneity Distribution and Other Statistics of the Estimated Models

	Without unobserved heterogeneity		With grade-constant unobserved heterogeneity		With grade-varying unobserved heterogeneity			
	Coeff.	Std. Err.	Coeff.	Std. Err.	Coeff.	Std. Err.		
<i>Unobserved heterogeneity probability masses</i>								
ρ_1	–	–	-0.8966	***	0.2632	-0.9840	***	0.1506
ρ_2	–	–	-2.0524	***	0.2319	-1.8887	***	0.2416
ρ_3	–	–	0.0000		–	0.0000		–
<i>Resulting probability masses</i>								
p_1		–		0.2655			0.2451	
p_2		–		0.0836			0.0992	
p_3		–		0.6509			0.6557	
Log-likelihood		-17,353.8		-17,281.2			-17,197.7	
AIC/ N		8.8715		8.8387			8.8003	
Number of parameters		92		100			108	
Number of pupils (N)		3,933		3,933			3,933	

Notes: *** Significant at 1%.

D.2 Estimation Results of the Equation for the Track Choice at the Beginning of Secondary School

Table A-2: Estimation results of the Track Choice at the Beginning of Secondary school

Variable	Without unobserved heterogeneity			With grade-constant unobserved heterogeneity			With grade-varying unobserved heterogeneity		
	Coeff.		Std. Err.	Coeff.		Std. Err.	Coeff.		Std. Err.
Calendar day of birth/100	-0.1649	***	0.0301	-0.1854	***	0.0346	-0.1890	***	0.0359
Years of delay at start of secondary school	-1.1880	***	0.1481	-1.3618	***	0.1806	-1.3514	***	0.1818
Female	0.2435	***	0.0607	0.2957	***	0.0722	0.3122	***	0.0743
Cohort 1980	-0.1526	**	0.0605	-0.1793	**	0.0717	-0.1960	***	0.0740
Father's education/10	1.6128	***	0.1147	1.8874	***	0.1506	1.9675	***	0.1486
Mother's education/10	1.7222		0.1261	2.0513	***	0.1611	2.0902	***	0.1614
<i>Number of siblings – Reference: No siblings</i>									
1 sibling	-0.1357		0.0949	-0.1732		0.1128	-0.1930	*	0.1160
2 siblings	-0.2776	***	0.1040	-0.3425	***	0.1236	-0.3789	***	0.1265
3 or more	-0.3890	***	0.1166	-0.4997	***	0.1384	-0.5775	***	0.1426
<i>Ordered logit thresholds</i>									
$\alpha_{1,tr}$	-3.6181	***	0.1269	-5.7932	***	0.3749	-6.4144	***	0.3414
$\ln(\alpha_{2,tr} - \alpha_{1,tr})$	0.5734	***	0.0426	0.6883	***	0.0495	0.6832	***	0.0447
$\ln(\alpha_{3,tr} - \alpha_{2,tr})$	-0.4637	***	0.0532	-0.3141	***	0.0610	-0.3055	***	0.0563
$\ln(\alpha_{4,tr} - \alpha_{3,tr})$	0.7777	***	0.0239	0.9536	***	0.0421	1.0060	***	0.0356
Unobs. heter. loading factor δ_{tr}	–		–	0.7523	***	0.1157	0.5506	***	0.1585

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

D.3 Estimation Results of the Evaluation Equation

Table A-3: Estimation results of the Evaluation Equation

Variable	Without unobserved heterogeneity			With grade-constant unobserved heterogeneity			With grade-varying unobserved heterogeneity		
	Coeff.		Std. Err.	Coeff.		Std. Err.	Coeff.		Std. Err.
<i>Time-constant variables</i>									
Years of delay at start of secondary school	-0.3600	***	0.1060	-0.7880	***	0.1690	-0.6881	***	0.1446
Female	0.4551	***	0.0507	0.6161	***	0.0709	0.5618	***	0.0668
Cohort 1980	-0.0504		0.0489	-0.0892		0.0666	-0.0908		0.0640
Father's education/10	0.0945		0.0908	0.5796	***	0.1489	0.4895	***	0.1311
Mother's education/10	0.2739	***	0.0978	0.8463	***	0.1609	0.7113	***	0.1431
<i>Number of siblings – Reference: No siblings</i>									
1 sibling	0.0783		0.0736	0.0423		0.1021	0.0380		0.0982
2 siblings	-0.0813		0.0796	-0.1748		0.1107	-0.1781	*	0.1056
3 or more	-0.0994		0.0921	-0.2634	**	0.1287	-0.2750	**	0.1218
<i>Ordered logit thresholds</i>									
$\alpha_{1,ev}$	-4.9890	***	0.1743	-9.2160	***	0.5382	-11.0402	***	1.4219
$\ln(\alpha_{2,ev} - \alpha_{1,ev})$	0.3102	***	0.0364	0.3768	***	0.0370	0.3527	***	0.0377
<i>Time-varying variables</i>									
<i>Track in year t – Reference: VHS</i>									
GHS+	0.4539	***	0.1649	-1.5122	***	0.3648	-1.1482	***	0.2955
GHS-	-1.0408	***	0.1176	-2.4244	***	0.2475	-2.0643	***	0.2092
THS+	-1.4480	***	0.1261	-2.4171	***	0.2074	-2.1034	***	0.1901
THS-	-1.3699	***	0.1202	-2.0922	***	0.1686	-1.8558	***	0.1606
<i>Grade in year t – Reference: Grade 7</i>									
Grade 8	-0.7613	***	0.1034	-0.9216	***	0.1088	-4.6753	***	1.5362
Grade 9 [§]	-0.4126	***	0.1128	-0.7146	***	0.1241	∞		-
Grade 10	-0.5370	***	0.1052	-0.9725	***	0.1244	-4.1727	***	1.5712
Grade 11	-0.4702	***	0.1060	-1.0158	***	0.1311	3.1741	***	3.7086
Grade 12 if VHS	-0.7217	***	0.2767	-1.3160	***	0.2708	-5.1179	***	1.5406
Last grade	0.5885	***	0.1343	-0.0732		0.1544	-4.1320	***	1.5060
<i>Downgrade at the end of year t-1 – Reference: No downgrade</i>									
1-step downgrade	0.1928	*	0.1134	0.1139		0.1267	0.0690		0.1262
2-step downgrade	0.2482		0.1965	-0.0691		0.2120	0.0348		0.2079
Ever retained before year t-1 (ψ_{ev})	-1.0267	***	0.1797	0.3986		0.2459	0.2149		0.3469
Retention at the end of year t-1 (κ_{ev})	-0.1544		0.2713	-3.2002	***	0.8451	-2.4716	*	1.4758
Heterogeneous effect of Ever retained before year t-1 (ψ_{ev}^0)	-0.2132	*	0.1199	0.0779	**	0.0350	0.0574		0.0428
Heterogeneous effect of Retention at the end of year t-1 (κ_{ev}^0)	-0.2741	**	0.1346	-0.6598	***	0.1191	-0.4127	**	0.1697
<i>Unobserved heterogeneity</i>									
<i>Unobserved heterogeneity support points – v_1^1 normalized to 0</i>									
$v_1^2(7)$	-		-	-4.6092	***	0.4885	-6.5837	***	1.5676
<i>Grade varying unobserved heterogeneity of $v_1^2(\cdot) - v_1^2(7)$</i>									
$v_1^2(8) - v_1^2(7)$	-		-	-		-	4.3095	***	1.6636
$v_1^2(9) - v_1^2(7)$	-		-	-		-	-15.7057	***	0.2898
$v_1^2(10) - v_1^2(7)$	-		-	-		-	3.4792	**	1.6762
$v_1^2(11) - v_1^2(7)$	-		-	-		-	-4.2500		3.7719
$v_1^2(12) - v_1^2(7) = v_1^2(13) - v_1^2(7)$	-		-	-		-	3.7905	**	1.6374
$v_1^3(7)$	-		-	-2.6645	***	0.4084	-4.7825	***	1.4038
<i>Grade varying unobserved heterogeneity of $v_1^3(\cdot) - v_1^3(7)$</i>									
$v_1^3(8) - v_1^3(7)$	-		-	-		-	3.7260	**	1.5041
$v_1^3(9) - v_1^3(7)$	-		-	-		-	-15.7728	***	0.1573
$v_1^3(10) - v_1^3(7)$	-		-	-		-	3.2926	**	1.5546
$v_1^3(11) - v_1^3(7)$	-		-	-		-	-4.0617		3.7004
$v_1^3(12) - v_1^3(7) = v_1^3(13) - v_1^3(7)$	-		-	-		-	4.6492	***	1.5177

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

[§] When we include grade-varying heterogeneity we need to fix the coefficient of grade 9 since it tends to be a very large number.

D.4 Estimation Results of the Drop-Out Equation

Table A-4: Estimation results of the School Drop-Out Equation

Variable	Without unobserved heterogeneity		With grade-constant unobserved heterogeneity		With grade-varying unobserved heterogeneity				
	Coeff.	Std. Err.	Coeff.	Std. Err.	Coeff.	Std. Err.			
<i>Time-constant variables</i>									
Years of delay at start of secondary school	-0.6436	***	0.2414	-0.7978	***	0.2580	-0.1486	0.2457	
Female	-0.2316		0.1665	-0.1992		0.1751	-0.1592	0.1699	
Cohort 1980	0.2335		0.1631	0.2776		0.1709	0.2157	0.1628	
Father's education/10	-0.8076	**	0.3139	-0.7160	**	0.3311	-0.7298	**	0.3168
Mother's education/10	-0.2614		0.3192	-0.0661		0.3526	-0.1525	0.3241	
Presence of siblings	-0.1530		0.2205	-0.1882		0.2290	-0.0403	0.2227	
Constant α_{out}	-0.3886		0.3863	0.8502		0.6724	1.0376	**	0.4711
<i>Time-varying variables</i>									
Track in year t : VHS [§]	2.8119	***	0.3325	3.3619	***	0.4105	2.3220	***	0.3809
Evaluation in year t : A [§]	-3.3873	***	0.3081	-3.9154	***	0.3505	-4.0952	***	0.3930
Grade in year t : final grade [§]	-2.4515	***	0.3686	-2.6123	***	0.3851	-3.3014	***	0.4246
1-step or 2-step downgrade at the end of year $t-1$ [§]	-1.0167	*	0.5632	-1.1814	*	0.6047	-0.5673		0.5819
Ever retained before year $t-1$ (ψ_{out}^0)	-0.7461	***	0.2317	-1.0981	***	0.3106	-0.6281	**	0.2955
Retention at the end of year $t-1$ (κ_{out})	-0.2450		0.3428	-0.6272		0.4678	0.5842		0.5577
Heterogeneous effect of Ever retained before year $t-1$ (ψ_{et}^0)	-0.2991	***	0.0835	-0.3436	***	0.0752	-0.2811	***	0.0802
Heterogeneous effect of Retention at the end of year $t-1$ (κ_{out}^0)	-0.2174	*	0.1160	-0.2652	***	0.1006	-0.1460		0.1097
Unobserved heter. loading factor δ_{out}	-		-	-0.4101	**	0.1918	-0.3104	**	0.1308 b)

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

[§] We had to group track, evaluation, grade, and track downgrade dummies into broader categories due to the scarce number of observations in some categories if defined at a finer level.

D.5 Estimation Results of the Resitting Equation for B Students

Table A-5: Estimation results of the Resitting Equation for B Students

Variable	Without unobserved heterogeneity		With grade-constant unobserved heterogeneity		With grade-varying unobserved heterogeneity		
	Coeff.	Std. Err.	Coeff.	Std. Err.	Coeff.	Std. Err.	
<i>Time-constant variables</i>							
Years of delay at start of secondary school	-0.5316	0.4986	-0.6861	0.5090	-0.8854	*	0.5251
Female	-0.1148	0.1854	-0.0674	0.1933	-0.0162		0.1971
Cohort 1980	0.2018	0.1870	0.2362	0.1895	0.2666		0.1988
Father's education/10	0.7340	** 0.3286	0.8598	** 0.3471	1.0480	***	0.3624
Mother's education/10	0.0190	0.3595	0.1461	0.3941	0.3033		0.3916
Presence of siblings	0.5060	* 0.2750	0.5825	** 0.2777	0.6222	**	0.2918
Constant α_{re}	-1.8411	*** 0.3380	-1.0999	0.8302	-0.5092		0.4950
<i>Time-varying variables</i>							
<i>Track in year t - Reference: THS+/THS-</i> [§]							
GHS+/GHS-	-0.3410	* 0.1995	-0.6003	*** 0.2326	-0.8633	***	0.2258
<i>Grade in year t - Reference: Grade 8</i>							
Grade 9 [†]	0.5879	** 0.2527	0.4328	* 0.2584	$+\infty$		-
Grade 10	1.1516	*** 0.2139	0.9546	*** 0.2275	1.1661	**	0.5300
1-step or 2-step downgrade at the end of year $t-1$ [§]	0.4993	0.3332	0.3788	0.3453	0.4466		0.3574
Ever retained before year $t-1$ (ψ_{re})	-0.3644	0.4318	-0.2809	0.4683	-0.1598		0.4779
Retention at the end of year $t-1$ (κ_{re})	-2.5382	** 1.1435	-2.4785	** 1.1477	-2.3282	**	1.1602
Unobserved heter. loading factor δ_{re}	-	-	-0.1732	0.2305	-0.7488	***	0.0601

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

[§] We had to group track and track downgrade dummies into broader categories due to the scarce number of observations in some categories if defined at a finer level.

[†] When we include grade-varying heterogeneity we need to fix the coefficient of grade 9 since it tends to be a very large number.

D.6 Estimation Results of the Track Downgrade Equation for GHS/THS Students

Table A-6: Estimation Results of the Track Downgrade Equation for GHS/THS Students

Variable	Without unobserved heterogeneity		With grade-constant unobserved heterogeneity		With grade-varying unobserved heterogeneity				
	Coeff.	Std. Err.	Coeff.	Std. Err.	Coeff.	Std. Err.			
<i>Time-constant variables</i>									
Years of delay at start of secondary school	0.2481	*	0.1463	0.3019	*	0.1565	0.3287	**	0.1542
Female	-0.0441		0.0634	-0.0576		0.0645	-0.0642		0.0641
Cohort 1980	-0.0594		0.0633	-0.0564		0.0633	-0.0523		0.0633
Father's education/10	-0.6019	***	0.1173	-0.6557	***	0.1297	-0.6942	***	0.1248
Mother's education/10	-0.3596	***	0.1280	-0.4287	***	0.1391	-0.4771	***	0.1338
<i>Number of siblings – Reference: No siblings</i>									
1 sibling	-0.0291		0.0933	-0.0255		0.0932	-0.0165		0.0931
2 siblings	-0.1076		0.1035	-0.0943		0.1041	-0.0740		0.1039
3 or more	-0.0038		0.1214	0.0157		0.1228	0.0458		0.1217
<i>Ordered logit thresholds</i>									
$\alpha_{1,dow}$	0.9913	***	0.1817	1.3774	***	0.3767	1.4731	***	0.2748
$\ln(\alpha_{2,dow} - \alpha_{1,dow})$	0.4950	***	0.0445	0.4951	***	0.0446	0.4980	***	0.0446
<i>Time-varying variables</i>									
<i>Evaluation and retention in year t – Reference: C</i>									
A	-1.6592	***	0.1229	-1.5529	***	0.1507	-1.4820	***	0.1354
B and not resitting	1.4570	***	0.1616	1.4644	***	0.1620	1.4703	***	0.1622
B and resitting	-0.8837	***	0.2424	-0.8585	***	0.2454	-0.7789	***	0.2456
<i>Track in year t – Reference: VHS–</i>									
GHS+	1.7198	***	0.1314	1.8787	***	0.1972	2.0138	***	0.1694
GHS–	0.8603	***	0.1221	0.9307	***	0.1415	0.9722	***	0.1308
THS+	1.2203	***	0.1334	1.2430	***	0.1367	1.2386	***	0.1348
<i>Grade in year t – Reference: Grade 8</i>									
Grade 7	-3.8673	***	0.2111	-3.8833	***	0.2121	-4.3862	***	0.4195
Grade 9	-1.2577	***	0.0917	-1.2454	***	0.0919	-4.3991	**	1.7462
Grade 10	-0.5479	***	0.0701	-0.5231	***	0.0732	-0.5625	***	0.1080
<i>Downgrade at the end of year $t-1$ – Reference: No downgrade</i>									
1-step downgrade	-0.4251	**	0.1878	-0.4203	**	0.1880	-0.3795	**	0.1862
2-step downgrade	-0.6925	***	0.4591	-0.6551	***	0.4628	-0.6368	***	0.4571
Ever retained before year $t-1$ (ψ_{dow})	0.7733	***	0.2362	0.5827	**	0.2693	0.5275	**	0.2364
Retention at the end of year $t-1$ (κ_{dow})	0.7582	***	0.1817	0.6013	***	0.1982	0.5883	***	0.1812
Heterogeneous effect of Ever retained before year $t-1$ (ψ_{dow}^0)	0.2884	*	0.1585	0.3037	*	0.1624	0.3500	**	0.1721
Heterogeneous effect of Retention at the end of year $t-1$ (κ_{dow}^0)	0.1772		0.1235	0.1778		0.1261	0.2380	*	0.1370
Unobserved heter. loading factor δ_{dow}	–		–	-0.0998		0.0820	-0.1787	**	0.0907

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

E Estimation Results if Downgrading Is First Decided upon by Students Getting a B with Grade-Varying Unobserved Heterogeneity

In the benchmark model we assume that the choice to repeat the grade in case of a B precedes the track downgrading decision. However, one might question this assumption: B students might first decide whether to stay in the current track and, if B students decide to remain in the same track, no choice about resitting is left. In this section, we report the estimation results of the model in which B students first are assumed to decide first whether to stay in the current track. After the estimation of this alternative model, which is non-nested to the benchmark one, we discriminate between models on the basis of the Vuong test for strictly non-nested models (Vuong, 1989, p. 316–319). We find that the alternative order of events could be rejected against the one of the benchmark model. The value of the asymptotically standard Normal statistic is 3.798 in favor of the benchmark model and rejects the alternative hypothesis at a p -value of 0.0001.

E.1 Estimated Probability Masses of the Discrete Unobserved Heterogeneity Distribution if Downgrading Is First Decided upon by Students Getting a B

Table A-7: Estimated Probability Masses of the Discrete Unobserved Heterogeneity Distribution and Other Statistics of the Model if Downgrading Is First Decided upon by Students Getting a B

	With grade-varying unobserved heterogeneity	
	Coeff.	Std. Err.
<i>Unobserved heterogeneity probability masses</i>		
ρ_1	-0.9822 ***	0.1531
ρ_2	-1.8634 ***	0.2429
ρ_3	0.0000	–
<i>Resulting probability masses</i>		
p_1		0.2448
p_2		0.1014
p_3		0.6538
Log-likelihood	-17,227.9	
AIC/ N	8.8166	
Number of parameters	110	
Number of pupils (N)	3,933	

Notes: *** Significant at 1%.

E.2 Estimation Results of the Equation for the Track Choice at the Beginning of Secondary school if Downgrading Is First Decided upon by Students Getting a B

Table A-8: Estimation results of the Track Choice at the Beginning of Secondary School if Downgrading Is First Decided upon by Students Getting a B

Variable	With grade-varying unobserved heterogeneity		
	Coeff.		Std. Err.
Calendar day of birth/100	-0.1889	***	0.0359
Years of delay at start of secondary school	-1.3520	***	0.1820
Female	0.3121	***	0.0742
Cohort 1980	-0.1965	***	0.0739
Father's education/10	1.9724	***	0.1485
Mother's education/10	2.0891	***	0.1614
<i>Number of siblings – Reference: No siblings</i>			
1 sibling	-0.1920	*	0.1160
2 siblings	-0.3815	***	0.1264
3 or more	-0.5817	***	0.1426
<i>Ordered logit thresholds</i>			
$\alpha_{1,tr}$	-6.4252	***	0.3407
$\ln(\alpha_{2,tr} - \alpha_{1,tr})$	0.6847	***	0.0445
$\ln(\alpha_{3,tr} - \alpha_{2,tr})$	-0.3084	***	0.0563
$\ln(\alpha_{4,tr} - \alpha_{3,tr})$	1.0066	***	0.0357
Unobserved heterogeneity loading factor δ_{tr}	0.5516	***	0.1611

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

E.3 Estimation Results of the Evaluation Equation if Downgrading Is First Decided upon by Students Getting a B

Table A-9: Estimation results of the Evaluation Equation if Downgrading Is First Decided upon by Students Getting a B

Variable	With grade-varying unobserved heterogeneity	
	Coeff.	Std. Err.
<i>Time-constant variables</i>		
Years of delay at start of secondary school	-0.6840	*** 0.1458
Female	0.5561	*** 0.0668
Cohort 1980	-0.0932	0.0638
Father's education/10	0.4847	*** 0.1318
Mother's education/10	0.7079	*** 0.1440
<i>Number of siblings – Reference: No siblings</i>		
1 sibling	0.0351	0.0982
2 siblings	-0.1816	* 0.1057
3 or more	-0.2830	** 0.1211
<i>Ordered logit thresholds</i>		
$\alpha_{1,ev}$	-11.0443	*** 1.4468
$\ln(\alpha_{2,ev} - \alpha_{1,ev})$	0.3534	*** 0.0376
<i>Time-varying variables</i>		
<i>Track in year t – Reference: VHS</i>		
GHS+	-1.1388	*** 0.3034
GHS–	-2.0550	*** 0.2138
THS+	-2.0905	*** 0.1930
THS–	-1.8585	*** 0.1631
<i>Grade in year t – Reference: Grade 7</i>		
Grade 8	-4.6766	*** 1.5674
Grade 9 [§]	12.6568	11.5651
Grade 10	-4.2873	*** 1.5783
Grade 11	2.0600	4.1883
Grade 12 if VHS	-5.1805	*** 1.5653
Last grade	-4.1756	*** 1.5322
<i>Downgrade at the end of year t–1 – Reference: No downgrade</i>		
1-step downgrade	0.0854	0.1266
2-step downgrade	0.0342	0.2092
Ever retained before year t–1 (ψ_{ev})	0.1795	0.3558
Retention at the end of year t–1 (κ_{ev})	-2.5926	* 1.4655
Heterogeneous effect of Ever retained before year t–1 (ψ_{ev}^0)	0.0561	0.0438
Heterogeneous effect of Retention at the end of year t–1 (κ_{ev}^0)	-0.4265	** 0.1667
<i>Unobserved heterogeneity</i>		
<i>Unobserved heterogeneity support points – v_1^1 normalized to 0</i>		
$v_1^2(7)$	-6.5628	*** 1.6000
<i>Grade varying unobserved heterogeneity of $v_1^2(\cdot) - v_1^2(7)$</i>		
$v_1^2(8) - v_1^2(7)$	4.3357	** 1.7106
$v_1^2(9) - v_1^2(7)$	-13.3026	11.5656
$v_1^2(10) - v_1^2(7)$	3.5516	** 1.6996
$v_1^2(11) - v_1^2(7)$	-3.1056	4.2206
$v_1^2(12) - v_1^2(7) = v_1^2(13) - v_1^2(7)$	3.8670	** 1.6791
$v_1^3(7)$	-4.7856	*** 1.4293
<i>Grade varying unobserved heterogeneity of $v_1^3(\cdot) - v_1^3(7)$</i>		
$v_1^3(8) - v_1^3(7)$	3.7178	** 1.5299
$v_1^3(9) - v_1^3(7)$	-13.3307	11.5615
$v_1^3(10) - v_1^3(7)$	3.4381	** 1.5571
$v_1^3(11) - v_1^3(7)$	-2.9476	4.1802
$v_1^3(12) - v_1^3(7) = v_1^3(13) - v_1^3(7)$	4.7205	*** 1.5438

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

E.4 Estimation Results of the Drop-Out Equation if Downgrading Is First Decided upon by Students Getting a B

Table A-10: Estimation results of the School Drop-Out Equation if Downgrading Is First Decided upon by Students Getting a B

Variable	With grade-varying unobserved heterogeneity	
	Coeff.	Std. Err.
<i>Time-constant variables</i>		
Years of delay at start of secondary school	-0.8741	** 0.3648
Female	-0.1486	0.2471
Cohort 1980	-0.1586	0.1709
Father's education/10	0.2145	0.1635
Mother's education/10	-0.7144	** 0.3228
Presence of siblings	-0.1345	0.3272
Constant α_{out}	-0.5751	0.5805
<i>Time-varying variables</i>		
Track in year t : THS [§]	-0.1465	0.1100
Evaluation in year t : A [§]	2.3716	*** 0.4148
Grade in year t : final grade [§]	-4.1158	*** 0.4299
1-step or 2-step downgrade at the end of year $t-1$ [§]	-3.2722	*** 0.4253
Ever retained before year $t-1$ (ψ_{out})	-0.0370	0.2237
Retention at the end of year $t-1$ (κ_{out})	-0.6018	** 0.2992
Heterogeneous effect of Ever retained before year $t-1$ (ψ_{ev}^0)	0.6099	0.5710
Heterogeneous effect of Retention at the end of year $t-1$ (κ_{out}^0)	-0.2801	*** 0.0806
Unobserved heter. loading factor δ_{out}	1.0168	** 0.4847

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

[§] We had to group track, evaluation, grade, and track downgrade dummies into broader categories due to the scarce number of observations in some categories if defined at a finer level.

E.5 Estimation Results of the Resitting Equation for B Students if Downgrading Is First Decided upon by Students Getting a B

Table A-11: Estimation results of the Resitting Equation for B Students if Downgrading Is First Decided upon by Students Getting a B

Variable	With grade-varying unobserved heterogeneity	
	Coeff.	Std. Err.
<i>Time-constant variables</i>		
Years of delay at start of secondary school	-0.9396 *	0.5253
Female	0.1797	0.2121
Cohort 1980	0.3592 *	0.2125
Father's education/10	0.8866 **	0.3962
Mother's education/10	0.5750	0.4209
Presence of siblings	0.8268 ***	0.3112
Constant α_{re}	-2.3301 ***	0.4192
<i>Time-varying variables</i>		
<i>Track in year t – Reference: THS+/THS–[§]</i>		
GHS+/GHS–	-0.4920 *	0.2609
<i>Grade in year t – Reference: Grade 8</i>		
Grade 9	15.3336	11.9525
Grade 10	1.1072 *	0.5965
1-step or 2-step downgrade at the end of year $t-1$ [§]	0.6850 *	0.3773
Ever retained before year $t-1$ (ψ_{re})	0.1626	0.5271
Retention at the end of year $t-1$ (κ_{re})	-2.0478 *	1.1930
Unobserved heter. loading factor δ_{re}	-0.5561	0.6561

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

[§] We had to group track and track downgrade dummies into broader categories due to the scarce number of observations in some categories if defined at a finer level.

E.6 Estimation Results of the Track Downgrade Equation for GHS/THS Students if Downgrading Is First Decided upon by Students Getting a B

Table A-12: Estimation Results of the Track Downgrade Equation for GHS/THS Students if Downgrading Is First Decided upon by Students Getting a B

Variable	With grade-varying unobserved heterogeneity		
	Coeff.		Std. Err.
<i>Time-constant variables</i>			
Years of delay at start of secondary school	0.3778	**	0.1560
Female	-0.0642		0.0639
Cohort 1980	-0.0564		0.0630
Father's education/10	-0.7433	***	0.1252
Mother's education/10	-0.4734	***	0.1338
<i>Number of siblings – Reference: No siblings</i>			
1 sibling	-0.0178		0.0932
2 siblings	-0.0673		0.1038
3 or more	0.0368		0.1208
<i>Ordered logit thresholds</i>			
$\alpha_{1,dow}$	1.6551	***	0.2888
$\ln(\alpha_{2,dow} - \alpha_{1,dow})$	0.4117	***	0.0457
<i>Time-varying variables</i>			
<i>Evaluation and retention in year t – Reference: C</i>			
A	-1.4662	***	0.1373
B	0.6607	***	0.1483
<i>Track in year t – Reference: VHS–</i>			
GHS+	2.1744	***	0.1741
GHS–	1.1213	***	0.1307
THS+	1.3436	***	0.1344
<i>Grade in year t – Reference: Grade 8</i>			
Grade 7	-4.5312	***	0.4713
Grade 9	-4.5630	*	2.6630
Grade 10	-0.5725	***	0.1149
<i>Downgrade at the end of year t–1 – Reference: No downgrade</i>			
1-step downgrade	-0.4150	**	0.1789
2-step downgrade	-0.5763		0.4540
Ever retained before year t–1 (ψ_{dow})	0.4727	**	0.2187
Retention at the end of year t–1 (κ_{dow})	0.6569	***	0.2055
Heterogeneous effect of Ever retained before year t–1 (ψ_{dow}^0)	0.3098	*	0.1819
Heterogeneous effect of Retention at the end of year t–1 (κ_{dow}^0)	0.3723	**	0.1572
Unobserved heter. loading factor δ_{dow}	-0.2141	**	0.0980

Notes: *** Significant at 1%; ** significant at 5%; * significant at 10%.

F Additional ATTs Based on Counterfactual Simulations

Table A-13: Treatment Heterogeneity of Grade Repetition (C) Relative to Forced Downgrading (B) after Grade 8

Treatment: C versus B evaluation after grade 8	Model: grade-varying unobserved heterogeneity		
	ATT		95% CI
<i>I. Evaluation in grade 9: A</i>			
First quartile	-0.194	**	[-0.384, -0.014]
Second quartile	-0.038		[-0.186, 0.096]
Third quartile	0.019		[-0.097, 0.123]
Fourth quartile	0.064		[-0.039, 0.155]
<i>II. High school graduation</i>			
First quartile	-0.179	*	[-0.371, 0.000]
Second quartile	-0.033		[-0.208, 0.128]
Third quartile	0.024		[-0.131, 0.159]
Fourth quartile	-0.007		[-0.181, 0.145]
<i>III. Delay at start last compulsory year</i>			
First quartile	0.874		[0.651, 1.106]
Second quartile	0.721	**	[0.538, 0.918]
Third quartile	0.659	***	[0.485, 0.844]
Fourth quartile	0.621	***	[0.458, 0.801]

Notes: All statistics are based on 999 random simulations of the treated sample that allow for the uncertainty of the estimated parameters. The ATTs are calculated by subtracting the average outcome in case of the counterfactual of forced downgrading from the average outcome in case of a retention. The first quartile is the one with the lowest value for the linear index of the evaluation in grade 8. ***, **, * indicate whether the ATT is significantly different from 0 (1) in panels I and II (panel III) at the 1%, 5%, 10% significance levels, respectively.

Table A-14: ATTs of Grade Repetition (C) Relative to Forced Downgrading (B) after Grade 9 and 10

	Model: grade-varying unobserved heterogeneity			
	Treatment: C versus B evaluation after grade 9		Treatment: C versus A evaluation after grade 10	
	ATT	95% CI	ATT	95% CI
<i>I. Evaluation in next grade: A</i>				
All treated	-0.055	[-0.149, 0.038]	–	–
<i>II. High school diploma</i>				
All treated	-0.038	[-0.128, 0.048]	-0.035	[-0.111, 0.042]
<i>III. Delay at start last compulsory year</i>				
All treated	0.643	*** [0.504, 0.767]	0.558	*** [0.440, 0.667]

Notes: All statistics are based on 999 random simulations of the treated sample that allow for the uncertainty of the estimated parameters. The ATTs are calculated by subtracting the average outcome in case of the counterfactual of forced downgrading from the average outcome in case of a retention. Panel *I* is empty for treatments in grade 10 since not all individuals reach grade 11 (and therefore outcomes for the latter year cannot be calculated for all individuals). ***, **, * indicate whether the ATT is significantly different from 0 (1) in panels *I* and *II* (panel *III*) at the 1%, 5%, 10% significance levels, respectively.

References

- Billingsley, P. (1995) *Probability and measure*, 3rd ed. (New York: Wiley)
- Carneiro, P., K. Hansen, and J.J. Heckman (2003) 'Estimating distributions of treatment effects with an application to the returns to schooling and measurement of the effects of uncertainty on college choice.' *International Economic Review* 44(2), 361–422
- Gaure, S., K. Røed, and T. Zhang (2007) 'Time and causality: A Monte Carlo assessment of the timing-of-events approach.' *Journal of Econometrics* 141(2), 1159–1195
- Mroz, T., G. Picone, F. Sloan, and A. Y. Yashkin (2016) 'Screening for a chronic disease: A multiple stage duration model with partial observability.' *International Economic Review* 57(3), 915–934
- Vuong, Q.H. (1989) 'Likelihood ratio tests for model selection and non-nested hypotheses.' *Econometrica* 57(2), 307–333