

Intergenerational transmission of language capital and economic outcomes

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Abstract: This paper investigates the intergenerational transmission of language capital in immigrant communities from one generation to the next, and the effect of language deficiencies on the economic performance of second generation immigrants. Our analysis is based on a long panel that oversamples immigrants and allows their children to be followed even after they have left the parental home. Our results show a significant and sizeable association between parental language fluency and that of their children, conditional on a rich set of parental and family background characteristics. We also find that language deficiencies of the children of immigrants are associated with poorer labour market outcomes for females, but not for males. Finally, we find a strong relationship between parental language fluency and labour market outcomes for females, which works through the child's language proficiency.

Keywords: Intergenerational transmission, Human Capital, Language Proficiency of Immigrants

JEL-Classification: J15, J24, J62

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1. Introduction

A large and growing literature is concerned with the intergenerational transmission of income and wealth. The process by which wealth is transmitted from one generation to the next is an important component in understanding inequality and its evolution over time, as emphasized in work by Becker and Tomes (1986). Recent empirical work finds sizeable differences in intergenerational mobility across countries, with studies for the US and the UK (see e.g. Solon 1992; Zimmerman 1992; Dearden, Machin and Reed 1997) establishing higher income immobility than studies for European countries like Sweden and Germany (see e.g. Bjorklund and Jantti 1997 and Wiegand 1997).

But even within the same country, the intergenerational transmission of income differs across populations. Immigrants and their children are an important subgroup which highlights this. Work by Borjas (1993) and Hammarstedt and Palme (2004) illustrates substantial differences in the intergenerational earnings correlation between different immigrant groups. There are several reasons why immigrants should exhibit different patterns of intergenerational transmission. Borjas (1992) emphasizes that the socioeconomic performance of the next generation depends not only on parental skills, but also on the ethnic skills of the parent's generation. Borjas (1995) argues that part of this ethnic capital may be due to selection of immigrants into particular neighbourhoods - a point that has been re-emphasised in work by Nielsen, Rosholm, Smith and Husted (2001) and Rooth and Ekberg (2003).

This paper investigates one particular aspect of human capital where parental endowment may affect transmission to the next generation: language capital. Language proficiency has long been understood to be a key factor in the process of economic assimilation of immigrants and their labour market performance. Earlier papers (see e.g. Carliner 1981, McManus, Gould and Welch 1983, Chiswick 1991, Dustmann 1994, Chiswick and Miller 1995, Shields and Wheatley-Price 2002) find a strong and significant effect of language proficiency on earnings and other economic outcomes. Work by Dustmann and van Soest (2001) shows that simple regression analysis may underestimate this effect due to measurement error – a conclusion that is supported by evidence reported in Bleakley and Chin (2004a). Hardly any work exists on how language proficiency affects educational and labour market outcomes of second generation immigrants.

Language is often cited as the principle initial barrier confronting recent immigrants (see e.g. Portes and Rumbaut 1996). As language is significant in determining economic outcomes, it is important to assess the extent to which parental language proficiency affects future generations and is transmitted from one generation to the next. There are many reasons to believe that language proficiency of second generation immigrants is related to the language proficiency of their parents. Chiswick, Lee and Miller (2005) investigate what determines the parent/child proficiency relationship. Children of immigrants may experience a monolingual environment in the home country language in the parental home, thus hindering the development of fluency in the host country language. Lack of exposure to a correct form of the host country language at early stages of the child's life may have long term consequences, affecting the child's entire educational chain and accumulation of human capital, and more directly, her labour market opportunities. It is a well known fact among cognitive scientists that languages are learnt more easily at very young ages (see e.g. Johnson and Newport 1989). Thus, parental proficiency

during the child's formative years in the home may be a critical determinant of the child's host-country language fluency level.

Using a long panel that oversamples immigrants and that allows their children to be followed even after they have left the parental home, we analyse the intergenerational transmission of language capital, and examine how language proficiency of second generation immigrants affects their labour market outcomes. We contribute to the literature in several ways. First, we study the effect of parental characteristics, and in particular parental proficiency in the host country language, on the language proficiency of their children. Bleakley and Chin (2004b) analyse the relationship between parents' language proficiency and that of their children for the US. Their analysis is based on data from the 2000 US Census which provides self-reported language proficiency of parents as well as their children. Chiswick, Lee and Miller (2005) analyse the language correlation between parents and their children using the 1996 Australian Census. Our data is quite different. Unlike Bleakley and Chin (2004b) and Chiswick et al. (2005), we have repeated information on both parents and their children. This allows us to address the problem of measurement error - which is serious in self-reported data on language (see Dustmann and van Soest 2001) - by using a measure which exploits the repeated information on language proficiency available to us. Also, while Bleakley and Chin (2004b) and Chiswick et al. (2005) observe children only before the age of 17, and if they have not yet left the parental household, our information on children's language proficiency is collected after the age of 16, and independent of the child leaving the parental household. This avoids selection and allows us to investigate not alone the association between parental language proficiency and that of their children, but also the children's later economic outcomes. Our sample is based on survey

data and accordingly smaller than the samples used in Bleakley and Chin (2004b) and Chiswick et al. (2005).

Parental language proficiency may be correlated with their child's language proficiency for reasons like education, transmission of ability, and cultural attitudes. We make use of the rich family background information in our data and condition on variables like parental education, age and origin as well as permanent parental income to eliminate or reduce channels other than language exposure at childhood that may lead to parental language proficiency being correlated with that of their child. As our data allows us to follow immigrant children even after they have left the parental household, our analysis is to our knowledge the first to investigate the effect deficiencies in language proficiency of second generation immigrants have on their economic outcomes (we investigate earnings, labour force participation, employment and unemployment), and how this relates to the language proficiency of their parents.

Our analysis distinguishes between males and females, and the children of immigrants who are born in the host country, and those who are born abroad, but arrived in the host country before the age of 10. Our results show a significant and sizeable association between parental language fluency and that of their children. Language deficiencies of the children of immigrants are associated with poorer labour market outcomes for females, but not for males. For females, we establish a clear relationship between parental language fluency and labour market outcomes which works through their language proficiency.

The structure of the paper is as follows. In the next section we give some theoretical considerations and explain our empirical strategy. In section 3, we discuss our data and samples, and provide some descriptive statistics. Section 4 presents results on intergenerational transmission of language proficiency, and section 5 analyses how language deficiencies of

second generation immigrants affect their labour market outcomes. Section 6 discusses our findings and concludes.

2. The transmission of language capital

Theoretical considerations

In the intergenerational permanent income model with parental investment and perfect capital markets (see Becker and Tomes 1986), earnings regress to the mean according to the intergenerational correlation of ability. Capital market imperfections lead to less rapid regression to the mean than through intergenerational mobility in endowments in families where credit constraints are binding. This framework is a useful starting point for empirical work, and Solon (1999, 2004) provides a structural interpretation of coefficients frequently estimated in studies of intergenerational income mobility based on the Becker-Tomes model.

It is less clear however that this framework is equally appropriate as a model for the study of other types of characteristics that are transmitted across generations, like language capital. As language proficiency is a part of human capital, part of any correlation between parental language proficiency and that of the child may be explained by the Becker-Tomes model, to the extent that parental language capacity is correlated with parental permanent earnings, and as far as intergenerational correlation in ability may result in a correlation between parent's and children's language proficiency. However, most of the correlation between children's and parent's language proficiency is likely to be driven by exposure to the host country language at earlier stages in the child's lifecycle. As suggested by the cognitive psychology literature, there

is a strong relationship between the age of exposure to a foreign language and later proficiency in it.²

There are a number of further factors that could be picked up by any correlation of language proficiency between parent and child. Parental language proficiency and that of their children may be related to the social and ethnic context in which children grow up. This argument is similar to that of Borjas (1992) who emphasizes the importance of ethnic capital for intergenerational mobility. Poor language capacity of the parent may also capture more intense ethnic networking, embedding the child into an environment where the host country language is not often used, and may seem to be of less value.

Bleakley and Chin (2004b) define the exposure effects of parental language proficiency on that of their children as their parameter of interest. To isolate these exposure effects, they use parental age of arrival as an instrument for parental language fluency, arguing that those parents who arrived at a young age learn the host country language more easily. As this variable may work on their children's language proficiency through channels other than parental language proficiency, they use for identification the interaction of this variable with non-English speaking country of origin. This identification strategy therefore assumes the same non-language related age of arrival effects for children of immigrants from English-speaking and non-English speaking countries.

² In the neurolinguistic literature, Penfield and Roberts (1959) proposed the biologically based "critical period" hypothesis for second language acquisition. Later studies (Johnson and Newport 1989, Birdsong and Molis 2001, Mayberry and Lock 2003) confirm the hypothesis that there exists a strong relationship between age at exposure to second language and later proficiency in it. Newport (2002) states that decline in average proficiency in second language acquisition can begin as early as ages 4-6. While there has also been some work (Hakuta K, Bialystok E, Wiley E. 2003) which disputes the critical period hypothesis, the consensus in the cognitive psychology literature appears to be that second language attainment is negatively correlated with age of learning.

Our approach makes use of the background characteristics that we have available in our survey data. Instead of using an instrumental variable type approach, we make a “selection on observables” assumption, by conditioning on different sets of factors that may lead to confounding the effect of parental language on the offspring’s language proficiency, and cultural and environmental factors (see Angrist and Krueger 1999 for discussion). The parameter we estimate is the change in the conditional expectation of the child’s language proficiency (L^c) if parental proficiency (L^p) changes, conditional on a set of background variables X : $\frac{\partial E(L^c | L^p, X)}{\partial L^p}$. To the extent that the set of conditioning variables eliminates the correlation of child’s language proficiency with parental language proficiency through any other confounding channels of the type we have discussed above, this measures the exposure effect. Notice that as our conditioning variables contain measures of parental permanent earnings, the mechanism that creates intergenerational mobility in the Becker – Tomes model is eliminated.

We implement this strategy by estimating regressions of the following form:

$$(1) \quad L_i^c = a_1 + \theta L_i^p + X_i' \alpha + v_i .$$

The variables L_i^c and L_i^p are measures of language proficiency, and we discuss their construction below. The vector of conditioning variables X_i includes family and background characteristics that take account of confounding factors of the sort discussed above. One such factor may be related to origin of the parent, as different home country languages may be more or less distant from the host country language. Origin dummies (measured as origin country of

the father³) also pick up ethnic capital (see Borjas 1992) and networking differences across groups. Correlation in genetic endowment may lead to a positive correlation between parental language proficiency and the child's language proficiency. We condition on parental education, as well as a permanent measure of father's earnings.⁴ Exposure of the child to the host country language may further depend on the extent to which parents are integrated in the host country society and labour market, as well as their social context. To capture this, we condition on the number of years the mother and the father have been in the host country when the child is aged 10 years, and survey information about contact of parents with residents of the host country.

Measurement of language proficiency

A key issue is measurement of the variables L_i^c and L_i^p . In our data, both are self-reported. Self-reported data on language proficiency suffer from measurement error. While this leads only to a loss in efficiency where the child's language measure is concerned, mismeasurement in parental language ability leads (as long as the measurement error is classical) to a downward bias in the parameter a_2 . Recent work by Dustmann and van Soest (2001) suggests that measurement error in language ability may lead to a downward bias by up to a factor of 3 in estimation equations that regress log wages on self reported language measures. This finding is supported by evidence provided in Bleakley and Chin (2004a).

We address the measurement error problem by making use of the repeated information we have on parental language proficiency to reduce the noise in our data. We construct a time-

³ In the 17 cases where fathers are absent, we use the origin country of the mother.

⁴ Again, in 28 cases where we have no earnings information on fathers, we use a permanent measure of mother's earnings. There are 17 children in our sample for which we have no earnings data for either father or mother.

averaged fixed measure of language proficiency for the individuals in our sample. We do this by estimating fixed effects language equations of the following form:

$$(2) \quad y_{it} = b_0 + age_i * b_1 + age_i^2 b_2 + u_i + e_{it},$$

where y_{it} is a measure of language proficiency for individual i in period t , age is the individual's age, e_{it} is an idiosyncratic error term, and u_i is an individual specific fixed effect.

Our measure for an individual's language proficiency is then the prediction $\hat{b}_0 + age * \hat{b}_1 + age^2 \hat{b}_2 + \hat{u}_i$, evaluated at the parent's age when the child was 10 years old.⁵ Age 10 is chosen here as this is the age that the secondary school track choice is made in Germany, and therefore a critical point in the child's future education and labour market outcomes⁶. We use the same estimation method to predict the language proficiency of the child (using child's age), where again we predict proficiency at age 10. Notice that choosing another age would not affect the estimates as it does not affect the variation in this variable *across* individuals. The coefficient estimates for b_0, b_1, b_2 will be unbiased and consistent as the number of individuals grows large. The estimate \hat{u}_i , though unbiased, will be consistent only as the number of periods grows large. Below, we will run robustness checks by increasing the minimum number of parental language observations on which we base estimation, therefore reducing the remaining measurement error.

⁵ A similar approach has been used by Dustmann and van Soest (2002).

⁶ At age 10, children (or their parents) have to choose between three different tracks: lower, intermediate, or higher secondary school. While lower and intermediate secondary school usually lead to apprenticeship training in blue or white collar professions, only higher secondary school allows access to university. See Dustmann (2004) for more details.

3. The data, the sample, and descriptive evidence

The data we use for this analysis stems from 19 waves (between 1984 and 2002) of the German Socio-Economic Panel (GSOEP), which is a household-based panel survey, similar to the PSID in the US or the BHPS in the UK. The GSOEP was initiated in 1984, when it over-sampled the then resident migrant population in West-Germany. In the first wave, about 4500 households with a German born household head, and an additional sample with about 1500 households with a foreign born household head were interviewed, and subsequent interviews took place on a yearly basis. The foreign born households were collected from the five largest immigrant communities at the time: Turkey, Spain, Italy, former Yugoslavia, and Greece. Importantly, questionnaires for these households are available in the home country language. The data is thus unique in providing repeated information on a large sample of immigrants over a long period of time.

From the foreign headed households, we construct a panel of 810 children which forms the basis of our analysis. These children come from 467 different households; 238 of these are one-child households – the structure of siblings in the household can be seen in Table A1 in the Appendix. Households in which either immigrant parent obtains German citizenship are excluded from our sample as this renders their children - 39 children in our case - ineligible for inclusion in the foreigner survey which contains the language proficiency questions. Households without children are also obviously excluded from our sample, as are households where the children are too young to complete an adult questionnaire in those years where language questions were asked.

Each individual in a relevant household and over the age of 15 is interviewed. The household head provides information about all other individuals in the household and those below the interviewing age. Individuals who leave households and form their own households are also tracked and included in the panel.

When individuals are 16 years old, they receive their own personal identification, and pointers to their mother and their father. We construct a sample of parent-child pairs. We follow all children in the sample after the age of 15, and construct a corresponding data set of all mothers and all fathers. We define a second generation immigrant as an individual who is born in Germany, and whose head of household is born abroad. We have 599 of these children in our sample. We also include children of foreign born parents who are themselves foreign born, but arrived before the age of 10. There are 211 of these children in our sample. Table 1 shows the distribution of the origin country of the head of household for the 810 children in the sample. As can be seen from this, the country from which the largest share of head of households in our sample originates is Turkey, both for children born in Germany and born abroad. Their mean age at arrival is 4.2 years, and Table A2 in the Appendix shows the distribution of arrival age. In most of our analysis, we distinguish between these two groups.

Detailed language information which we use in our analysis is reported in 11 waves of the GSOEP.⁷ Individuals are asked to report their fluency in German on a five point scale. The question is “*How well do you speak the German language...?*” and possible responses are: *Very well...Well...Satisfactory...Badly...Very badly*. We scale this information between 0 (for very badly) and 1 (for very well).⁸

⁷ Language data is contained in waves 1984 – 1987 and then every second year from 1987 – 2001.

⁸ The same weight is given to each change in the scale, i.e. the values that we attribute to the above 5 levels are 0 (very badly), 0.25 (badly), 0.5 (satisfactory), 0.75 (well), 1 (very well).

We have 810 children in our panel for whom we have language observations. We also observe their parents' language proficiency over several years. Tables 2 and 3 give more information on the frequency of this language data.

Table 2 displays the number of language observations we have for the children and their parents in our panel. We have multiple language observations for 79% of the children, with language proficiency observed just once for the remaining 21%. 98% of mothers and fathers also have multiple language observations.

Table 3 shows the percentage of children and parents in our sample that have language observations in each wave of the GSOEP that contains questions on language proficiency. As the children mature into the adult questionnaire (which contains the questions on language), the percentage of parents that remain in the panel to answer language questions decreases. However, this does not pose a problem for us, as it is parental fluency when the child is at a young impressionable age that is most relevant to our analysis.

The kernel densities of the predicted language proficiency of all the parents and children in our sample (by gender of the child) are displayed in Figure 1. The parents' predictions are obtained as explained above; those of their children are obtained in the same way, and fixed at the child's age of 10. As we explain above, we use age 10 as the reference age because the secondary school choice is made at that age, which is important for future careers. Again, the age at which we fix the prediction only affects the location of this distribution, but not its shape.

The left panel of the figure displays densities of the language proficiency of both mothers and fathers. Parental distributions are quite dispersed, with a clear gender difference. The mean and median for fathers is .537 and .532 respectively, and for mothers .428 and .432, with corresponding standard deviations of .138 for fathers and .190 for mothers. The right panel

displays corresponding distributions for the children. Although a large proportion of children are concentrated towards the upper part of the distribution, there is a sizeable fraction of the sample at the intermediate part of the unit language fluency scale.

In Table 4 we display information about children in the sample and their parents. The last column in each panel is the p-value on the difference of the respective characteristic between individuals born in Germany and abroad. Many mean characteristics of second generation children born in Germany and abroad as well as their parents are significantly different. German born second generation immigrants have more years of schooling than those born abroad, with the differential being larger for females than it is for males. The means for spoken language proficiency (predicted at the age of 10 years) show that those born in the host country are more proficient in the host country language than those born abroad. Labour force participation is higher amongst children born abroad, but in this case the differential is larger between males than females.

Fathers of children who are born in Germany are slightly older than fathers of children who are born abroad, and fathers are older than mothers. This difference is larger for children who are born in the host country. The years of residence (which are computed when the child was 10 years old) are higher for parents of children born in the host country, with a clear difference between males and females, which suggests the typical pattern of male migration and subsequent female migration.

Parental hourly earnings are log hourly permanent earnings of the father, or, where there is no data on father's earnings (28 cases), permanent log hourly earnings of the mother. The data provides information on average monthly gross earnings in the month preceding the interview, and on hours worked for pay during that month. From that information, we compute a log hourly

wage rate. We compute permanent log hourly earnings by running fixed effects regressions of log hourly earnings on the individual's age and its square (where the earnings are deflated by a CPI). Our measure of permanent log hourly earnings⁹ is the sum of the individual fixed effect and the age polynomial, weighted by the estimated coefficients and evaluated when the child was aged 10.

Permanent log hourly wages are higher amongst parents of children who are born in the host country, as is the percentage of mothers employed when the child is aged 16 years. This may partly be explained by the fact that parents of children born in the host country have been in Germany about 7 years longer and have slightly more years of education than parents of those born abroad.

4. Intergenerational Transmission of Language Capital

Basic estimates

Table 5 reports results of regressions of child's proficiency in the host country language on parental language proficiency. We distinguish between children who are born in Germany, and children who are born abroad but who enter the host country before the age of 10. The parental language measure is a linear combination with equal weights on father's and mother's language measures, which are computed according to equation (3) (based on estimation of models as in (2)), and normalised between 0 and 1.¹⁰ The table reports three different specifications for both

⁹ 87 percent of sample individuals report hourly earnings in at least 4 years, and 70 percent in at least 8 years.

¹⁰ There are 12 parent-child pairs where language information on the mother is missing, and 17 parent-child pairs where language information on the father is missing. In these cases, we use predicted language proficiency for the parent present as parental measure.

categories of children, with different sets of conditioning variables. The reported coefficients are the parental language measure, gender and cohort.

Columns (1) reports results where we regress the child's language measure on the parental language measure only. Columns (2) include in addition the child's cohort, a gender dummy, and the number of siblings. Finally, columns (3) add parental background variables, which include a self-reported measure for contact with other Germans¹¹, the years since the father's¹² migration when the child was aged 10 years old, mother's and father's years of education, and father's permanent log wage.¹³ We also experimented with larger sets of conditioning variables; these resulted in almost identical results as those reported in columns (3).¹⁴ The numbers in parenthesis beside the tick (\checkmark) are p-values for the joint significance of the respective set of regressors. The coefficients on parental language can be interpreted as elasticities, as both the child's and parent's language measures are scaled between 0 and 1.

The estimates across the different columns decrease when we add family background characteristics. The p-values indicate that the additional sets of conditioning variables are jointly significant. This suggests that some of the correlation between parental language fluency and that of their children is absorbed by the set of conditioning variables, in particular by parental background. However, the estimates we obtain remain large and significant, suggesting a strong association between parental language proficiency and that of their children. For children born in the host country, the estimated standard deviation of parental language proficiency is about 0.14, so a point estimate of 0.25 implies that a one-standard deviation increase of parental language

¹¹ This is a dummy variable if the parent answers yes to the question 'Have you had contact with Germans?' when their child was aged 10 or younger.

¹² Mother's years since migration when the child was aged 10 is used in the 30 cases where father's information on years since migration is missing.

¹³ Again, in 28 cases where we are missing wage information on the father, we use that of the mother.

¹⁴ Additional conditioning variables included mother's years since migration, mother's and father's age, mother's and father's school leaving degree, and an average measure of parental education.

proficiency leads to an increase in language proficiency of the child of approximately 3.5 percentage points. For children born abroad, the estimated effect is about 5.0 percentage points.

For children born in the host country, there is a strong gender effect, with males being more fluent than females. This effect, if anything, is reversed for children who are born abroad. Finally, as all these estimates – both for children born in Germany and born abroad – are less than one, this suggests the occurrence of a catch-up effect for children across generations.

Which parent is more important?

In our analysis above, we have regressed child's language proficiency on language measures of both parents. It is not unlikely that father's and mother's language ability affects the language proficiency of their offspring differently. In Table 6, we present estimates where we regress child's language separately on mother's and father's language and then on both mother's and father's language together.

The estimates refer to the specification which includes the full set of conditioning regressors (corresponding to columns (3) in Table 5). For children born in the host country, the effects of mother's and father's language proficiency seem to be similar in magnitude. Conditioning on both measures at the same time reveals that mother's proficiency is slightly more important than father's proficiency (although not significantly so), which may be related to children having more intensive contact with the mother in the household. For children born abroad, it seems that it is mainly father's proficiency that is associated with the child's language fluency.

Males and Females

In Table 7, we report estimates where we allow parental language fluency to have different effects on males and females, based on the most general specification in Table 5, where we condition on the full set of individual and parental background information. The upper panel of the table reports results for parental language, and the lower panel allows for different impacts of father's and mother's language.

The results in the upper panel suggest a larger impact of parental language fluency on females than on males. The coefficient is particularly large for females who are born abroad, and suggests that an increase in parental fluency by one standard deviation increases fluency of daughters who are born abroad by about 8 percentage points. These differences between males and females could be explained by the hypothesis that children have more exposure to mother's than father's language in the home, and as females are more likely to spend time in their mother's company than males, then parental language will have a larger impact on females than on males.

In the lower panel of the table, we allow for different effects of father's and mother's language proficiency on male and female immigrant children. For males born in Germany, father's and mother's proficiency has roughly equal impact, while for males born abroad, it seems to be mainly father's language proficiency that affects fluency of the son. For females, there appears to be a slightly stronger association between their fluency and that of the mother.

Robustness checks

As is typical in studies of intergenerational mobility using survey data of the type we use, many children in our sample have the same mother or father. In Table A1 in the Appendix we illustrate the sibling structure in our sample. More than 70 percent of all children in our data have a brother or sister who is also in the sample.¹⁵ To check whether this affects our estimates, we have re-estimated all the models above, restricting our sample by using only the oldest or only child within a foreign headed household. The estimates of the effects of parental language on child's language from these regressions are reported in Tables 8a and 8b (using the same specifications as in columns (3), Table 5 above). Although the standard errors are slightly increased due to the decrease in sample size, the point estimates are very similar to those reported in Tables 5 and 7, and they are all statistically significant.

As we discussed above, a particular concern in studies of intergenerational transmission is measurement error. The measures of parental language proficiency we use in our analysis are in most cases based on repeated information for the same individual. Table 2 shows that 97 percent of mothers and fathers have reported their language proficiency in at least three interviews. As the construction of our parental language measure makes use of all language information reported in the sample, this suggests that any downward bias due to measurement error is significantly reduced in the estimates we report. To check whether a more radical selection would affect our estimates, we re-estimate our models based on father-mother pairs where each partner reports at least five language spells. The estimates we obtain are very similar to those we report in Table 5 and we include them in the Appendix in Table A3.

¹⁵ We adjust standard errors in our estimates above to take account of clustering within families.

5. Language Proficiency and Labour Market Outcomes

Language fluency, wages, and employment

Having established a relationship between parental language proficiency and language proficiency of the child, we now turn to examining the effects of the child's language proficiency on their labour market outcomes as second generation immigrants. We investigate four outcomes: labour force participation, employment, unemployment, and wages. We describe the construction of these variables in Table 4 above.

We estimate models of the following type:

$$(3) \quad Y_{it}^c = b + \delta L_i^c + Z_{it}' \beta + u_{it}.$$

Here L_i^c is a measure of language proficiency, and we use as before the predicted language proficiency at age 10, as reported in Table 4. Z_{it} is a vector of conditioning variables, and $u_{it} = \lambda_i + e_{it}$ is an error term, with λ_i being an individual specific random effect. For each outcome Y_{it}^c , we estimate linear random effects models on all available observations, to take account of the covariance structure induced by repeated information on the same individual. Separate regressions are estimated for males and females, and also for those born in Germany and those born abroad. All regressions include only individuals who have finished full time education¹⁶. We report results in Table 9.

For each outcome we firstly report results (first row of each panel) which condition on educational attainment, age and its square, year dummies, age on arrival and its square for those

¹⁶ We drop children who are still in full time education as their labour market outcomes are not comparable to those who have completed their education.

children born abroad and number of children for females, and secondly, results which do not condition on education, but on family background information instead, where we add the same set of variables as in columns (3) of Table 5 (second row of each panel), as well as the number of children for females.¹⁷

The results suggest no significant effect of language proficiency on any of the outcomes for males. For females however, the coefficient estimates are mostly significant, with estimates being similar for the two specifications. The parameter estimates are larger for those who are born abroad than for those who are born in the host country.

For wages, estimates are significant only for females who are born abroad. Here an increase in one standard deviation of language proficiency increases wages by about 13 percent, which is a sizeable effect.

Labour market participation for both females born in the host country as well as females born abroad is positively associated with language proficiency, with the coefficient estimate being larger for those who are born abroad. Conditional on educational achievements, an increase in language proficiency by one standard deviation increases the participation probability for females born in the host country by about 6 percentage points, while it increases the participation probability of females born abroad by about 14 percentage points.

Looking at employment, we again see that estimates for females born abroad are larger than they are for those born in the host country. An increase of one standard deviation in language proficiency improves employment probabilities by around 8 percentage points for females born

¹⁷ These include parental country of origin and cohort controls, and a full set of parental controls for parental years since migration and parental earnings when the child was 10 years old, both parents years of education, and parental contact with Germans when the child was 10 years old or younger (as in columns (3) in Table 4).

in the host country, but by up to 14 percentage points for females born abroad, when we condition on education. This suggests that for females born abroad, language proficiency may have to compensate for a lack of social networks which assist in obtaining employment; networks which females born in Germany may be able to take advantage of in their job search.

Estimation results for unemployment probabilities are similar. The probability of unemployment is significantly reduced by fluency – yet again, the effects are larger for those who are born abroad, with an increase in language proficiency by one standard deviation reducing unemployment risk by about 7 percentage points, conditional on education. Overall, the effects for females are sizeable, and seem to be larger for children of immigrants who are born abroad rather than in the host country. Again, the possibility that second generation immigrants who are born in the host country may find it easier to compensate for deficiencies in language fluency than immigrants who are born abroad could explain this finding.

These results suggest that fluency deficiencies in the host country language are detrimental for the labour market outcomes of second generation females, but not for males. One reason for this gender difference may be that females find employment opportunities predominantly in jobs where language fluency is quite important (for example, services), while males have more job opportunities in occupations where language is less important. Figures from the ILO (2003) suggest that in industrialised countries, females are more concentrated in the service sector, while males are more concentrated in industry. For Germany, 7 of the 10 most popular occupations for males (employing nearly 25 percent of all males) are either in crafts (e.g. car mechanic or carpenter) or manual work (e.g. warehouseman); for females, the 10 most popular occupations (employing 38 percent of all females) are all in services or white collar jobs (numbers are for 2000; own computations based on IABS administrative data). While females

with language deficiencies may in principle take up ‘male’ jobs that require less communicative and language skills, they may be disadvantaged in such jobs due to the physical activity involved or employer prejudice, thus making these jobs less accessible for them.¹⁸

Parental Fluency and Child’s Labour Market Outcomes

In the previous sections, we have computed the association between parental fluency and the fluency of the child, as well as the effect of the child’s fluency on their labour market outcomes. In this section, we assess the importance of parental language fluency on labour market outcomes of the child. To infer the effect of parental fluency on the child’s outcomes, we combine the effect of parental fluency on child’s fluency with the effect of child’s fluency on labour market performance, using the parameter estimates we have obtained from equations (1) and (3). Hence, we specify exactly the mechanism by which parental fluency affects the child’s labour market outcome. As this imposes restrictions on the way parental language affects the child’s outcomes, our estimates are more efficient than direct estimation obtained by regressing the child’s outcomes on parental language proficiency – the latter would be the reduced form effect, allowing for all other influences of parental language (other than only through the child’s language).

The intergenerational language equation we estimate is the same specification as in the upper panel of Table 7. The specifications which relate child’s labour market outcome to language are

¹⁸ In our sample, and classifying individuals who work into 9 occupation groups, about 52% of males work in the craft and related trades occupations, compared to only 7% of females. In contrast, 38% of females work in occupation groups that consist of jobs in the service sector, shops, and market related sales, while only 7% of males work in these groups.

the same as specifications II in Table 9. Accordingly, both specifications condition on the same extensive set of background variables. The effect of parental fluency on the child's labour market outcome, which works through its effect on child's language proficiency, is therefore the product of these two parameters. Denote the estimated parameter of parental language proficiency on child's language proficiency by $\hat{\theta}$, with standard error $\hat{\sigma}_\theta$, and the estimated parameter of child's language proficiency on child's labour market outcomes by $\hat{\delta}$, with standard error $\hat{\sigma}_\delta$. The estimates we present in Table 10 are the product $\hat{\theta}\hat{\delta}$.¹⁹

As we reported above, language fluency is not associated with labour market outcomes for males; we should therefore expect the effect of parental fluency on males which works through the child's language fluency, to be insignificant as well. For females born abroad, we find that the labour force participation probability is positively and significantly affected by parental language proficiency. An increase in parental fluency by one standard deviation increases participation by about 4.5 percentage points. For those born in the host country, an increase in parental language proficiency by one standard deviation, increases the participation probability by 1.7 percentage points and the employment probability by 2.3 percentage points. According to these estimates, language fluency of first generation immigrants has not only an immediate effects on labour market outcomes, as established in earlier work (see e.g. Chiswick and Miller 1995 and Dustmann and van Soest 2002) but, by way of affecting their proficiency in the host country language, impacts on the labour market opportunities of their female children.

¹⁹ Under the assumption that $E(u_{ii} | Z_{ii}, X_i, v_i) = 0$, the covariances between δ and θ are zero so that the standard error can be approximated by the delta method. The reported standard error is computed as

$$se(\hat{\theta}\hat{\delta}) = \sqrt{\hat{\theta}^2 \hat{\sigma}_\delta^2 + \hat{\delta}^2 \hat{\sigma}_\theta^2}.$$

6. Discussion and Conclusions

Language fluency of first generation immigrants is increasingly regarded as a key requirement for qualifying as an immigrant. Economic research has established a strong link between language proficiency of immigrants, and their labour market performance. This paper provides further evidence of the importance of language proficiency for first generation immigrants. It argues that poor language knowledge of immigrants may affect fluency of second generation immigrants, and that poor fluency may in turn adversely affect labour market performance of second generation immigrants. Thus, improvement of language fluency of first generation immigrants may not only have an immediate impact, but may also play a role in improving the performance of second generation immigrants.

To analyse the intergenerational link between language proficiency, and the effects of language proficiency on the outcomes of second generation immigrants requires data on parents' and their children's language fluency, as well as information about labour market outcomes of second generation immigrants after they have entered the labour market. We use a long panel for Germany which oversamples immigrants to obtain this information. We construct measures of language proficiency where measurement error is reduced by an averaging type procedure. We find that second generation immigrants are far more fluent in the host country language than their parents. However, a significant percentage still has language deficiencies. There is a strong and significant association between parental proficiency in the host country language and the proficiency of their children. This effect slightly decreases, but remains strong even after conditioning on a large set of family and parental background information. The association is larger for children who are born abroad but entered the host country before the age of 10, than it is for children who are born in the host country. It is also larger for females than it is for males.

We find sizeable effects of language fluency of second generation immigrants on their labour market outcomes for females, but not for males. One explanation is that males find it easier to switch to jobs where language proficiency is less important, while job opportunities for females are concentrated in sectors like services, where language proficiency is more important. We provide evidence that parental language proficiency - through its effect on the child's language fluency - has detrimental effects on employment and participation probabilities of females.

Our analysis is based on survey data, which is unique in the way that it contains an over sample of immigrants, and provides panel information on language fluency and labour market outcomes for both generations. Our results suggest that one reason for the poor labour market performance of second generation immigrants is fluency deficiencies in the host country language. Our findings emphasise the importance of fluency in the host country language for first generation immigrants, and add further weight to policies that ensure language proficiency. Transmission of language deficiencies is found to affect in particular females, and may therefore contribute to disadvantage female second generation immigrants, as compared to their male counterparts. The findings provide first evidence on a link between language deficiencies of immigrants and labour market disadvantage of their female children.

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Tables and Figures

Table 1: Child's Country of Birth by Head of Household's Country of Origin

<i>HH Country of Origin</i>	<i>Children Born Germany</i>	<i>Children Born Abroad</i>	Total
Turkey	222	111	333
Ex-Yugoslavia	120	43	163
Greece	91	15	106
Italy	113	32	145
Spain	53	10	63
Total	599	211	810

Source: GSOEP, all waves 1984 – 2002.

Table 2: Number of Times Language Observed for Children and Parents

# of Language Observations	0	1	2	3	4	5	6	7	8	9	10	11	Total
% Children	0 (0)	21.48 (174)	19.75 (160)	17.28 (140)	11.73 (95)	9.38 (76)	7.90 (64)	4.94 (40)	3.33 (27)	2.84 (23)	1.23 (10)	0.12 (1)	100 (810)
% Mothers	0.99 (8)	0.12 (1)	0.49 (4)	1.36 (11)	2.47 (20)	3.83 (31)	5.68 (46)	9.51 (77)	8.89 (72)	9.75 (79)	12.72 (103)	44.20 (358)	100 (810)
% Fathers	2.10 (17)	0 (0)	1.11 (9)	1.48 (12)	2.72 (22)	3.58 (29)	5.19 (42)	10.62 (86)	8.64 (70)	10.12 (82)	13.33 (108)	41.11 (333)	100 (810)

Source: GSOEP, all waves 1984 – 2002.

Note: number of observations in parentheses

Table 3: Number of Children's and Parent's Language Observations in Each Wave

Year	1984	1985	1986	1987	1989	1991	1993	1995	1997	1999	2001	Total
<i>Children</i>												
% Obs	0.25 (2)	7.28 (59)	15.06 (122)	22.35 (181)	33.70 (273)	45.56 (369)	52.22 (423)	48.02 (389)	46.54 (377)	43.70 (354)	39.88 (323)	2872
<i>Mothers</i>												
% Obs	95.29 (769)	95.29 (769)	94.42 (762)	96.28 (777)	95.04 (767)	91.95 (742)	85.75 (692)	77.82 (628)	67.53 (545)	58.49 (472)	52.54 (424)	7347
<i>Fathers</i>												
% Obs	97.73 (775)	96.97 (769)	94.96 (753)	96.47 (765)	94.45 (749)	91.42 (725)	85.50 (678)	75.79 (601)	66.46 (527)	56.49 (448)	49.43 (392)	7182

Source: GSOEP, all waves 1984 – 2002.

Note: number of observations in parentheses

Table 4: Sample Characteristics

<i>Children</i>	<i>Females</i>			<i>Males</i>		
	Born Germany	Born Abroad	p-values	Born Germany	Born Abroad	p-values
Age Arrival	--	4.51 (2.45)		--	3.98 (2.28)	
Years Education ¹	10.69 (1.97)	9.88 (1.56)	<i>0.001</i>	10.44 (2.07)	10.12 (1.68)	<i>0.199</i>
Earnings ²	2.20 (0.34)	2.09 (0.32)	<i>0.000</i>	2.35 (0.36)	2.36 (0.30)	<i>0.431</i>
% Unemployed*	8.85	12.40		9.82	9.11	
% Labour Force Part.*	56.32	60.12		69.71	79.16	
% Employed*	50.77	52.18		62.52	71.33	
Spoken German ³	0.84 (0.13)	0.75 (0.17)	<i>0.008</i>	0.90 (0.10)	0.73 (0.11)	<i>0.000</i>
Siblings ⁴	74.33	73.68	<i>0.674</i>	75.25	71.55	<i>0.591</i>
Sample size	300	95		299	116	
<i>Parents⁵</i>						
Mothers' Age	36.79 (6.10)	36.11 (5.61)	<i>0.313</i>	37.46 (6.26)	36.45 (5.71)	<i>0.118</i>
Fathers' Age	41.46 (6.31)	39.01 (5.09)	<i>0.000</i>	41.82 (6.16)	40.33 (5.22)	<i>0.015</i>
Mothers' YSM ⁶	14.65 (4.12)	7.49 (3.96)	<i>0.000</i>	14.96 (4.12)	8.00 (4.17)	<i>0.000</i>
Fathers' YSM	17.48 (4.63)	10.74 (3.91)	<i>0.000</i>	17.22 (4.23)	11.45 (4.52)	<i>0.000</i>
Parental YSM	17.27 (4.70)	10.73 (3.83)	<i>0.000</i>	17.13 (4.27)	11.34 (4.52)	<i>0.000</i>
Mothers' Yrs Education ⁷	8.93 (1.94)	8.28 (1.61)	<i>0.001</i>	8.85 (1.96)	8.10 (1.27)	<i>0.000</i>
Fathers' Yrs Education	9.73 (1.96)	9.19 (1.98)	<i>0.024</i>	9.60 (1.98)	9.43 (1.84)	<i>0.422</i>
Parental Earnings ⁸	2.43 (0.24)	2.28 (0.25)	<i>0.000</i>	2.44 (0.22)	2.34 (0.25)	<i>0.000</i>
Mothers' German	0.48 (0.18)	0.36 (0.19)	<i>0.000</i>	0.47 (0.18)	0.34 (0.18)	<i>0.000</i>
Fathers' German	0.56 (0.14)	0.51 (0.14)	<i>0.007</i>	0.54 (0.13)	0.51 (0.13)	<i>0.011</i>
Parental German	0.52 (0.14)	0.43 (0.15)	<i>0.000</i>	0.50 (0.14)	0.43 (0.13)	<i>0.000</i>
% Mother's Employed	55.17	41.76		52.41	36.94	
% Father's Employed	86.02	90.59		83.68	92.66	
Sample Size: Mothers	298	95		298	116	
Sample Size: Fathers	293	91		294	115	

Note: In the above table: the # in the 1st column is the mean of the variable in question, the # in parentheses refers to the standard deviation, and the # in italics is the p-value from t-tests of mean equality of the variables between females born in Germany and born abroad, and between males born in Germany and born abroad. Sample size: 810 Children of Immigrant Parents.

1: Mean years of education of immigrant children who have completed education.

2: Earnings are the mean log hourly earnings of immigrant children who have completed education.

3: Siblings refers to the % of children who have at least one sibling.

4: Spoken German is a measure of child's language proficiency, predicted at age 10 as explained earlier in the paper.

5: Parental variables are fixed when the child was 10 years old (Mothers'/Fathers' age, Mothers'/Fathers'/Parental years since migration, Parental log hourly earnings, Mothers'/Fathers'/Parental spoken German), or 16 years old (% Mother's/Father's employed).

6: Mothers'/Fathers'/Parental YSM refer to their years since migration.

7: Mothers'/Fathers' Years Education are the maximum years of education obtained by parents.

8: Parental Earnings are the permanent log hourly earnings of fathers (or in 28 cases where missing data on fathers, mothers' permanent log hourly earnings is used).

*: Labour Force Participation is as defined by the ILO. Those Unemployed are defined as all labour force participants who are not working. Those Employed are defined as all those in our sample who are working.

Table 5: OLS Language Regressions; Dependent variable is Child's Language Proficiency

	<i>Children Born Germany</i>			<i>Children Born Abroad</i>		
	(1)	(2)	(3)	(1)	(2)	(3)
Parental Language	0.301 (0.034)**	0.308 (0.036)**	0.252 (0.049)**	0.377 (0.072)**	0.373 (0.076)**	0.327 (0.110)**
Cohort		0.001 (0.001)	0.001 (0.002)		-0.001 (0.003)	0.008 (0.005)
Male		0.071 (0.009)**	0.073 (0.010)**		-0.026 (0.019)	-0.039 (0.018)*
<i>Controls for</i>						
Age at Arrival				√ (0.000)	√ (0.000)	√ (0.000)
Individual Background		√ (0.000)	√ (0.000)		√ (0.000)	√ (0.000)
Parental Background			√ (0.000)			√ (0.000)
Observations	599	599	556	211	211	175
R-squared	0.11	0.19	0.23	0.13	0.15	0.26

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%.

(1): No controls. (2): Controls for individual background – cohort, gender, siblings. (3): Controls for individual & parental background – cohort, gender, siblings, fathers' years of education, mothers' years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Table 6: OLS Language Regressions; Dependent variable is Child's Language Proficiency

	<i>Children Born Germany</i>			<i>Children Born Abroad</i>		
	(1)	(2)	(3)	(1)	(2)	(3)
Mother's Language	0.174 (0.039)**		0.145 (0.039)**	0.154 (0.084)		0.098 (0.084)
Father's Language		0.170 (0.045)**	0.108 (0.046)*		0.270 (0.096)**	0.230 (0.105)*
Observations	552	556	552	188	192	188
R-squared	0.22	0.21	0.23	0.19	0.21	0.22
P-values	0.00	0.00	0.00	0.00	0.00	0.00

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%.

(1): No controls. (2): Controls for individual background – cohort, gender, siblings. (3): Controls for individual & parental background – cohort, gender, siblings, fathers' years of education, mothers' years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Table 7: OLS Language Regressions; Dependent variable is Child’s Language Proficiency: Males and Females

	Males		Females	
	<i>Born Germany</i>	<i>Born Abroad</i>	<i>Born Germany</i>	<i>Born Abroad</i>
Parental Language	0.189 (0.058)**	0.214 (0.104)*	0.312 (0.060)**	0.523 (0.161)**
Observations	556	175	556	175
R-squared	0.24	0.29	0.24	0.29
Mother’s Language	0.091 (0.046)*	0.028 (0.088)	0.202 (0.053)**	0.201 (0.133)
Father’s Language	0.114 (0.057)*	0.223 (0.109)*	0.099 (0.064)	0.281 (0.164)
Observations	552	171	552	171
R-squared	0.24	0.29	0.24	0.29

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%.

All columns control for individual & parental background – cohort, gender, siblings, fathers’ years of education, mothers’ years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Table 8a: OLS Language Regressions; Dependent variable is Child’s Language proficiency. Using the Oldest or Only Child in the household.

	<i>Children Born Germany</i>			<i>Children Born Abroad</i>		
	(1)	(2)	(3)	(1)	(2)	(3)
Parental Language	0.298 (0.050)**	0.328 (0.051)**	0.268 (0.069)**	0.331 (0.080)**	0.298 (0.092)**	0.331 (0.152)*
Observations	225	225	213	118	118	104
R-squared	0.11	0.21	0.25	0.14	0.17	0.28
P-values	0.00	0.00	0.00	0.00	0.00	0.00

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%

(1): No controls. (2): Controls for individual background – cohort, gender, siblings. (3): Controls for individual & parental background – cohort, gender, siblings, fathers’ years of education, mothers’ years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Table 8b: OLS Language Regressions; Dependent variable is Child’s Language proficiency. Using the Oldest or Only child in the household: Males and Females.

	Males		Females	
	<i>Born Germany</i>	<i>Born Abroad</i>	<i>Born Germany</i>	<i>Born Abroad</i>
Parental Language	0.203 (0.095)*	0.153 (0.139)	0.329 (0.091)**	0.549 (0.204)**
Observations	213	104	213	104
R-squared	0.25	0.32	0.25	0.32

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%.

All columns control for individual background – cohort, gender, siblings. (3): Controls for individual & parental background – cohort, gender, siblings, fathers’ years of education, mothers’ years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Table 9: Effect of Child's Language on Labour Market Outcomes

	Males		Females	
	<i>Born Germany</i>	<i>Born Abroad</i>	<i>Born Germany</i>	<i>Born Abroad</i>
Wages				
I	-0.012 (0.254) <i>628</i>	0.076 (0.260) <i>430</i>	0.269 (0.222) <i>501</i>	0.749 (0.293)* <i>229</i>
II	0.289 (0.311) <i>565</i>	0.012 (0.300) <i>415</i>	0.069 (0.262) <i>484</i>	0.677 (0.479) <i>217</i>
Labour Market Participation				
I	0.006 (0.116) <i>1471</i>	-0.195 (0.147) <i>761</i>	0.467 (0.117)** <i>1300</i>	0.804 (0.193)** <i>607</i>
II	0.041 (0.123) <i>1347</i>	-0.253 (0.163) <i>738</i>	0.376 (0.136)** <i>1254</i>	0.584 (0.193)** <i>550</i>
Employment				
I	0.113 (0.166) <i>1475</i>	0.037 (0.233) <i>768</i>	0.602 (0.122)** <i>1311</i>	0.835 (0.231)** <i>610</i>
II	0.072 (0.176) <i>1351</i>	-0.051 (0.263) <i>745</i>	0.537 (0.136)** <i>1264</i>	0.640 (0.304)* <i>553</i>
Unemployment				
I	-0.087 (0.132) <i>1418</i>	-0.251 (0.186) <i>735</i>	-0.244 (0.089)** <i>1104</i>	-0.397 (0.271) <i>438</i>
II	-0.012 (0.151) <i>1297</i>	-0.234 (0.199) <i>713</i>	-0.278 (0.090)** <i>1071</i>	-0.356 (0.330) <i>398</i>

Note: standard errors in parentheses; sample size in italics; * significant at 5%; ** significant at 1%.

(I): Conditioning on child's education, child's age, number of children for females, time dummies and age arrived in Germany for children born abroad.

(II): Not conditioning on child's education; including parental controls – fathers' years of education, mothers' years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans, as well as controls for child's age, cohort, siblings, number of children for females and time dummies.

Table 10: Child's Labour Market Outcomes and Parental Language Proficiency, 2 Step Estimates

	Males		Females	
	<i>Born Germany</i>	<i>Born Abroad</i>	<i>Born Germany</i>	<i>Born Abroad</i>
Wages	-0.080 (0.060)	-0.159 (0.119)	-0.067 (0.089)	0.014 (0.300)
Labour Force Participation	-0.004 (0.030)	-0.045 (0.047)	0.123 (0.042)**	0.299 (0.139)**
Employment	0.002 (0.035)	0.003 (0.056)	0.164 (0.052)**	0.270 (0.163)
Unemployment	-0.001 (0.025)	-0.058 (0.052)	-0.092 (0.037)**	-0.075 (0.144)

Note: standard errors in parentheses; sample size in italics; * significant at 5%; ** significant at 1%.

These estimates are the combination of the coefficient on the child's language measure when regressed on their different labour market outcomes, and the coefficient on the parental language measure when regressed on the child's language measure. The standard errors are derived using the Delta method. Controls for both sets of regressions

include fathers' years of education, mothers' years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans, as well as controls for child's age, cohort, siblings and number of children for females.

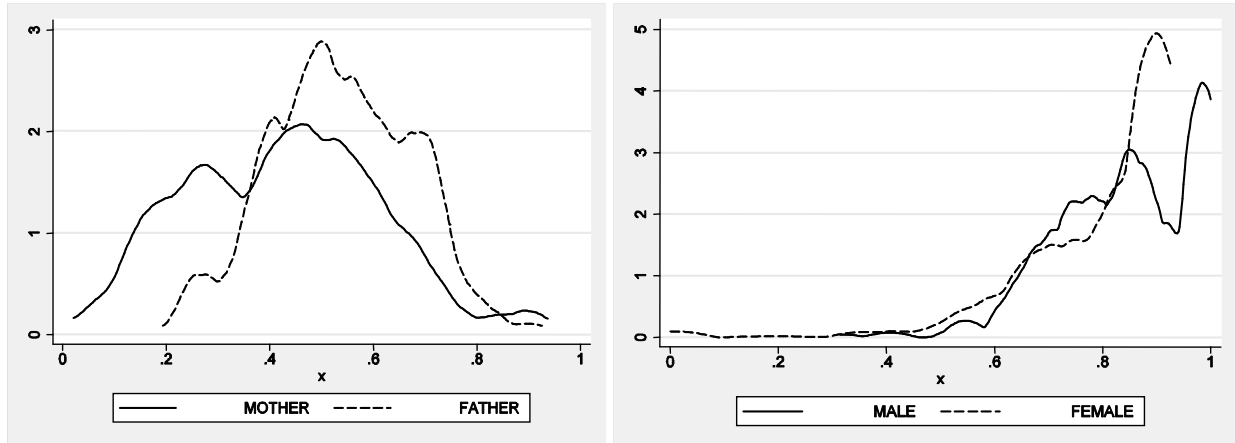


Figure 1: Kernel density of language proficiency, parents (left panel) and children (right panel)

Appendix

Table A1: Distribution of Siblings in Sample

# of Siblings in Family	# of Children	% of Children	# of Households	% of Households
0	238	29.38	238	50.96
1	292	36.05	146	31.26
2	177	21.85	59	12.63
3	76	9.38	19	4.07
4	20	2.47	4	0.86
5	0	0.00	0	0.00
6	7	0.86	1	0.21
Total	810	100.00	467	100

Source: GSOEP, all waves 1984 – 2002.

Table A2: Age at Arrival of Foreign Born Children

	Age Arrived in Germany									
	1	2	3	4	5	6	7	8	9	Total
% of Sample	16.39	12.67	16.39	14.89	7.32	14.59	5.40	7.98	4.38	100
# Children	29	21	33	30	14	26	13	12	10	188*

Source: GSOEP, all waves 1984 – 2002.

* Year of immigration is missing for 23 of foreign born children.

Table A3: OLS Language Regressions; Dependent variable is Child's Language Proficiency. Using only parents with 5 or more language observations.

	Children Born Germany			Children Born Abroad		
	(1)	(2)	(3)	(1)	(2)	(3)
Parental Language	0.289 (0.036)**	0.297 (0.038)**	0.251 (0.052)**	0.323 (0.087)**	0.317 (0.093)**	0.233 (0.111)*
Observations	556	556	530	180	180	160
R-squared	0.10	0.19	0.22	0.09	0.11	0.25
P-values	0.00	0.00	0.00	0.00	0.00	0.00

Source: GSOEP, all waves 1984 – 2002.

Note: Robust standard errors in parentheses; * significant at 5%; ** significant at 1%

(1): No controls. (2): Controls for individual background – cohort, gender, siblings. (3): Controls for individual & parental background – cohort, gender, siblings, fathers' years of education, mothers' years of education, parental years since migration, parental log hourly wages, country of origin of the head of household, age arrived in Germany for those children born abroad, parental contact with Germans.

Definition of variables used in regressions:

Parental Language: the average of mother's/father's spoken German proficiency when the child is 10 years old (or just mother's/father's proficiency if either is missing), predicted using a fixed effects model, scaled on the unit interval.

Age Arrived: the age of the child when they immigrated to Germany for those born abroad.

Head of Household's Country of Origin: the father's (or if missing, mother's) country of origin - Ex-Yugoslavia, Greece, Italy, and Spain, with Turkey being the reference country.

Siblings: the number of siblings that each child in the sample has; those with no siblings are the reference group.

Parental Years Since Migration: the father's (or if missing, mother's) years since migration when the child was aged 10 years old.

German Contact: dummy variable which is 1 if either parent had contact with Germans when the child was 10 years or younger.

Parental Earnings: the father's (or if missing mother's) log hourly wage predicted using a fixed effects model, evaluated when the child was aged 10 years old.

Mother's/Father's Education: the maximum years of education obtained by each parent.