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Educational Attainment and Risk Preference.

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Abstract:

We explore the relationship between risk preference and educational attainment for a sample of adults drawn from the 1996 U.S. *Panel Study of Income Dynamics (PSID)*. Using a sequence of questions from the 1996 *PSID*, we construct measures of an individual's risk aversion and risk tolerance allowing us to explore the implications of interpersonal differences in risk preference for educational attainment. Our empirical findings suggest that an individual's degree of risk aversion (tolerance) is inversely (positively) associated with their educational attainment. In addition, using the 1997 and 2002 *Child Development Supplements* of the *PSID*, we explore the relationship between the risk preference of parents and the academic achievements of their children. Our findings suggest that a parent's degree of risk aversion (tolerance) is negatively (positively) related to the academic achievements of their children.

Key Words: Human Capital; Risk Aversion; Risk Preference

JEL Classification: J24; J30

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I. Introduction and Background

Given the uncertainty surrounding returns to investments in human capital, it is not surprising that the risk preference of individuals has played a key role in the theory of human capital accumulation.¹ By definition, any investment in human capital can be considered risky, since the return is unknown and uncertain. For example, Palacios-Huerta (2003) finds that, due to the degree of risk associated with human capital investments, the actual gains from higher education per unit of risk in the U.S. are in the region of 5 to 20 per cent higher than that from risky financial assets.² Furthermore, it is not clear how one can reduce the degree of risk associated with human capital investments. As pointed out by Shaw (1996), the standard approach to reducing risk in financial investment, namely diversification, is often not available in the context of human capital investment. Typically, an individual holds one job with his/her human capital investments tailored accordingly. Hence, given the risk associated with returns to human capital investments, as well as difficulties with the diversification of such investments, the risk preference of individuals plays an important role in the decision to acquire human capital.

Given the obvious problems in measuring individuals' risk preferences, it is not surprising that attitudes towards risk have attracted limited attention in the empirical literature. In some empirical models of human capital accumulation, a parameter of constant risk aversion has been included,³ but such an approach does not allow variation in risk preferences across individuals to play a role in the investment decision-making process. Belzil and Hansen (2004), for example, estimate a dynamic programming model of schooling decisions where the degree of risk aversion is inferred from school decisions. In this model,

¹ See, for example, Johnson (1978), Levhari and Weiss (1974) and Gibbons and Murphy (1992).

² Harmon *et al.* (2003a) have adjusted the returns to schooling for individual risk by estimating Mincerian wage equations allowing for random coefficients, which yields dispersion (i.e. risk) in the returns to schooling by assigning individual specific returns.

³ Such studies include Brown and Rosen (1987), Moore (1987) and Murphy and Topel (1987).

individuals are assumed to be heterogeneous with respect to ability yet homogenous with respect to the degree of risk aversion.

An important exception in the literature is Shaw (1996) who jointly models investment in risky human capital and financial wealth allowing for interpersonal differences in risk preference. Shaw (1996) presents a theoretical framework which predicts an inverse relationship between an individual's degree of risk aversion and investment in risky human capital. The model is based on a portfolio allocation framework extended to incorporate an individual's decision to invest in risky human capital. Since human capital accumulation is modeled as a standard investment process, the less risk averse individuals are predicted to invest in relatively high levels of education. The empirical analysis suggests that risk preference affects the returns to human capital, although the relationship between risk preference and educational attainment is not directly explored. Brown and Taylor (2005) find supporting evidence using British panel data. In a similar vein, Brunello (2002) presents a theoretical framework which predicts that risk aversion affects educational choice via the marginal utility of schooling. The theoretical framework predicts that selected years of schooling decrease when absolute risk aversion increases. The empirical findings, which are based on Italian household survey data, support the theoretical priors. This result has received further recent empirical support from Guiso and Paiello (2007). Belzil and Leonardi (2007) also use Italian survey data to explore whether the transition from different levels of education changes with risk aversion and parental education background. Their findings suggest that different attitudes towards risk do not determine the level of schooling.

In a similar vein, Barsky *et al.* (1997) present measures of preference parameters relating to risk tolerance, time preference and inter-temporal substitution based on the U.S. *Health and Retirement Study*. The authors explore how risk preference varies across individual characteristics and report a 'U' shaped relationship between years of education

completed and risk tolerance. Individuals with twelve years of schooling were found to be the least risk tolerant, whilst individuals with more than sixteen years of schooling were found to have greater than average risk tolerance. In the multivariate regression analysis, however, the findings suggest that years of schooling are not associated with risk preference. It should be acknowledged that intuitively one might argue that a risk averse individual may have an incentive to invest heavily in human capital in order to safe-guard his/her future. Belzil and Hansen (2004) find that a counterfactual increase in risk aversion increases educational attainment, i.e. human capital accumulation.

In sum, the relationship between risk preference and educational attainment has attracted attention in both the empirical and the theoretical literature on human capital accumulation. According to such arguments, educational attainment (i.e. human capital accumulation) is influenced by risk preference. Our paper contributes to this area – specifically we further explore the relationship between risk preference and human capital accumulation from an empirical perspective exploiting a measure a risk preference elicited from individuals’ responses to a hypothetical gamble. In addition, we explore the relationship between a parent’s risk preference and the academic achievements of their offspring. Given the important role that parents play in decisions regarding their children’s education, such an intergenerational link, which to our knowledge has not attracted attention in the previous empirical literature, may unveil an additional determinant of children’s educational attainment.

II. Data

The obvious problem with exploring the relationship between human capital and risk preference from an empirical perspective lies in locating a suitable measure of risk preference. For this purpose, we exploit the *Panel Study of Income Dynamics (PSID)*, which is a representative panel of individuals ongoing since 1968 conducted at the Institute for Social

Research, University of Michigan. The *PSID* 1996 Survey includes a Risk Aversion Section, which contains detailed information on individuals' attitudes towards risk. The Risk Aversion Section contains five questions related to hypothetical gambles with respect to lifetime income.

To be specific, all heads of household were asked the following question (M1): *Suppose you had a job that guaranteed you income for life equal to your current total income. And that job was (your/your family's) only source of income. Then you are given the opportunity to take a new, and equally good, job with a 50-50 chance that it will double your income and spending power. But there is a 50-50 chance that it will cut your income and spending power by a third. Would you take the new job?*⁴ The individuals who answered 'yes' to this question, were then asked (M2): *Now, suppose the chances were 50-50 that the new job would double your (family) income, and 50-50 that it would cut it in half. Would you still take the job?* Those individuals who answered 'yes' to this question were then asked (M5): *Now, suppose that the chances were 50-50 that the new job would double your (family) income, and 50-50 that it would cut it by 75%. Would you still take the new job?* Individuals who answered 'no' to Question M1 were asked (M3): *Now, suppose the chances were 50-50 that the new job would double your (family) income, and 50-50 that it would cut it by 20 percent. Then would you take the job?* Those individuals who replied 'no' were asked (M4): *Now, suppose that the chances were 50-50 that the new job would double your (family) income, and 50-50 that it would cut it by 10 percent. Then would you take the new job?*

We use the responses to this series of questions to create a six point risk aversion index, RA_i , as follows (the percentages of individuals in each category are also shown below):

⁴ As Luoh and Stafford (2005) point out it is important to acknowledge that the question states that the new job will be 'equally as good' such that there is no difference in the non monetary characteristics of the jobs. Without such a qualification, individuals may be less willing to accept the gamble if there are non monetary attachments to their current job (Barsky *et al.*, 1997).

$RA_i =$	0 if $M1 = Yes \ \& \ M2 = Yes \ \& \ M5 = Yes$	5.88%
	1 if $M1 = Yes \ \& \ M2 = Yes \ \& \ M5 = No$	12.76%
	2 if $M1 = Yes \ \& \ M2 = No$	17.99%
	3 if $M1 = No \ \& \ M3 = Yes$	19.21%
	4 if $M1 = No \ \& \ M3 = No \ \& \ M4 = Yes$	20.72%
	5 if $M1 = No \ \& \ M3 = No \ \& \ M4 = No$	23.44%

Thus, the index is increasing in risk aversion such that if an individual rejects all the hypothetical gambles offered, the risk aversion index takes the highest value of 5, whilst if the individual accepts all gambles offered the risk aversion index takes the value of zero. It is interesting to note the low (high) percentage of respondents with the lowest (highest) value of the risk aversion index. Intermediate cases lie in between these two extreme values such that individuals are ranked according to their reluctance to accept the hypothetical gambles. The series of questions, thus, enables us to place individuals into one of six categories of risk aversion. Furthermore, as stated by Barsky *et al.* (1997), who find that this risk tolerance measure does predict risky behaviour such as smoking, drinking alcohol, not having insurance, choosing risky employment and holding risky financial assets, ‘the categories can be ranked by risk aversion without having to assume a particular form for the utility function,’ p.540.

A further measure of risk preference is available from the *PSID* 1996 survey based on Questions M1 to M5 above. Based on Barsky *et al.* (1997), a measure of risk tolerance (RT_i) is available in the *PSID* where the answers to Questions M1 to M5 have been converted into a single quantitative index of risk tolerance, which has been corrected for measurement error.⁵ Thus, RT_i being a measure of risk tolerance is inversely related to risk aversion.⁶

⁵ A detailed explanation of the conversion procedure is given by Luoh and Stafford (2005), which is summarized here. The risk tolerance data are taken from the last column of Table 1 in Barsky *et al.* (1997). Assume a utility function, $U(c) = (1/(1-1/q))c^{1-1/q}$, that q is log-normally distributed and $G = \ln(q)$. We observe G^* which lies in one of the categories determined by the hypothetical gamble questions. The product of each individual’s probability of being in a particular category yields the likelihood function. Maximizing the likelihood function and computing expected means conditional on being in a particular category yields q (Luoh and Stafford, 2005).

III. Educational and Risk preference

Our sample is restricted to those heads of household in employment in 1996 aged between 18 and 65, yielding a total of 5,277 observations.⁷ We explore the relationship between risk preference and human capital accumulation by modeling education, e_i , as a function of risk preference:

$$e_i = f(\mathbf{X}_i, r_i) + \varepsilon_{i1} \quad (1)$$

where r_i denotes the measure of risk preference and \mathbf{X}_i represents a set of additional explanatory variables, which draws on Wilson *et al.* (2005) and includes: age; gender; ethnicity; the mothers' marital status when the respondent was born; whether the respondent lived with his/her parents until age 16; whether the parents worked when the respondent was growing up; fathers' occupational status when the respondent was growing up; number of siblings; whether the respondent was the first born; whether the mother was born outside of the U.S.; the educational attainment of both parents; type of religion and whether the family was poor when the respondent was growing up.

In order to ascertain the robustness of our findings, we analyze both measures of risk preference, RA_i and RT_i . Similarly, we explore two measures of education – an index denoting the highest educational attainment of the head of household (e_i) and the number of years of completed schooling by the head of household (s_i). The highest educational attainment variable is a five point index where: 0 denotes less than high school completed, i.e. less than grade 12; 1 denotes high school completed; 2 denotes that the individual went to college but did not graduate;⁸ 3 denotes that the individual graduated from college; and, finally, 4 denotes that the individual completed some postgraduate education. The number of

⁶ It should be re-iterated that our measures of risk preference are based on hypothetical rather than actual behavior. In Section IV, we explore an alternative measure of risk preference based on actual behavior.

⁷ We focus on employees given the nature of the risk aversion question in the *PSID*, which relates to income from employment, i.e. income from employment is explicitly stated as the only income source.

⁸ This category includes individuals who went to vocational schools.

years of completed schooling is a continuous variable with a minimum (maximum) of 8 (17) years of schooling. We estimate equation (1) as an ordered probit model when measuring education by the index denoting the highest educational attainment of the head of household (e_i) given the inherent ordering of the index. When measuring education using the number of years of schooling completed by the head of household (s_i), equation (1) is estimated by ordinary least squares (OLS).

Table 1 presents a correlation matrix; between RA_i and RT_i ; between e_i and s_i ; and between the risk preference and educational attainment measures. The degree of correlation between the two measures of risk preference is in accordance with *a priori* expectations, i.e. the measure of risk tolerance is inversely related to the index of risk aversion. Moreover, the strong inverse relationship is significant at the one per cent level. Similarly, the correlation between e_i and s_i is positive and significant at the one per cent level. Finally, the measures of education are inversely associated with risk aversion and positively associated with risk tolerance. Such relationships between the key variables are in accordance with the empirical findings of Brown and Taylor (2005), Brunello (2002), Guiso and Paiella (2007) and Shaw (1996). Summary statistics relating to the variables used in our empirical analysis are presented in Table 2.

In Table 3 the results of estimating equation (1) are summarized for both measures of education: Panels A and B presents the results for the highest education attainment index (e_i) whilst Panel C presents the results for years of completed schooling (s_i). Due to the ordered nature of the highest education attainment index (e_i), we present the marginal effects of risk preference on the probability of having each level of education from no education, i.e. less than high school where index, e_i , equals zero, through to having completed some postgraduate study, where e_i , equals four. For each measure of education, we estimate two

specifications: specification 1 includes RA_i in the set of explanatory variables whilst specification 2 includes RT_i in the set of explanatory variables. The set of explanatory variables in each of these models is as shown in Table 4, which gives the full estimation results of the educational attainment equations where risk preference is measured by RA_i .⁹

It is apparent that, for our sample of 5,277 individuals, there is a statistically significant association between risk preference and both measures of education. The marginal effects show that risk preference, as measured by the risk aversion index (RA_i), is associated with an increase in the probability of having less than high school education (Table 3 Panel A). Indeed, a one standard deviation increase in the risk aversion index is associated with an increase in the probability of having less than high school education by 1.9%.¹⁰ Similarly, at the opposite end of the educational attainment hierarchy, a one standard deviation increase in risk aversion is associated with a decrease in the probability of having completed postgraduate study by 1%. Consistent results are found with the alternative measure of educational attainment, shown in Table 3 Panel C, where increasing risk aversion is inversely associated with the number of years of completed schooling (s_i). A one point move up the risk aversion index is associated with a decrease in the number of years of completed schooling by around 1.9%.¹¹ We also consider the effect of risk tolerance (RT_i) upon both measures of education. As expected, risk tolerance is inversely associated with the probability of having less than high school education (Table 3 Panel B) and correspondingly positively related to the number of years of schooling (Table 3 Panel C). For example, a one standard

⁹ In accordance with the existing literature we find that educational attainment is increasing in: age; father's occupation; mother's and father's educational attainment; religion; whether the individual is male and whether the individual was the firstborn. Factors which significantly decrease educational attainment are ethnicity and the number of siblings.

¹⁰ These calculations are based on the mean sample characteristics of individuals. For example, the 1.9% effect is calculated by multiplying the marginal effect, 0.0118, by the standard deviation of the risk aversion index, 1.6314.

¹¹ This is calculated as a risk preference elasticity such that $(\partial s / \partial r) \times (\bar{r} / \bar{s})$, where $\hat{\phi} = (\partial s / \partial r)$ and \bar{r} and \bar{s} denote the mean values of r and s respectively.

deviation increase in risk tolerance is associated with a decrease in the probability of having education less than high school by 1.6% and an increase in the probability of having completed some postgraduate study by 0.9%. Noticeably, risk preference generally has a monotonic relationship with educational attainment. In general, our findings accord with those of Brunello (2002) and Guiso and Paiella (2007) in that risk aversion is found to be inversely associated with educational attainment.¹²

IV. Risk Preference and Time Invariance

A potential problem with our measures of risk preference is that, as argued by Brunello (2002), educational choice depends upon risk attitudes at the time of the choice rather than risk preference observed in 1996. Hence, exploring the relationship between the 1996 risk preference measures and educational attainment may be problematic as human capital investments may have been made some time ago, for instance, when the individual was at high school. The inclusion of the 1996 risk preference measures in the educational attainment equation may be appropriate if risk preferences do not vary over time or if the human capital investments were made in 1996. The extent of the time variance issue may depend on the gap between the age of the individual in 1996 and the age of the individual when the human capital investment was undertaken. This potential problem was pointed out by Brunello (2002), but was not explicitly addressed in his empirical analysis of Italian survey data. Thus, the measures of risk preference are only meaningful if risk preferences are time invariant or if the influence of the time variant component of risk preference is small in terms of magnitude. In order to explore such issues, researchers have analyzed the relationship between age and proxies for risk preference such as the propensity to hold risky financial assets. Based upon

¹² A related line of enquiry relates to whether risk preference influences the returns to human capital investment. If risk preferences influence educational attainment, which in turn influences earnings, it may be the case that omitting risk preference in an educational attainment model may bias any estimates of the returns to education. Following Brunello (2002), we have explored the validity of risk preference as an over-identifying instrument for education in an earnings function. In accordance with Brunello (2002), we find that risk preference is a valid instrument for education in a wage equation.

U.S. data, Haliassos and Bertaut (1995), for example, found the impact of an individual's age on the decision to hold risky stocks to be statistically insignificant. More recently, Guiso *et al.* (2003) have also found the share of assets held in risky stocks to be invariant with respect to age effects in a number of countries. Such findings suggest that risk preferences do not vary with age.

We explore this issue in three ways. Firstly, we analyse the significance and magnitude of age effects in the risk preference equation. Secondly, we estimate the educational attainment equations by age cohorts in order to ascertain the robustness of the influence of risk preference on education across time. Clearly, for the youngest cohort, the potential gap between 1996 and the age of the human capital accumulation, will, on average, be the smallest. Thirdly, we exploit information relating to risk preferences from previous waves of the *PSID*.

We firstly model our measures of risk preference (r_i) conditional upon: age; age squared; ethnicity; gender; marital status; household size, number of children in the household; and whether the individual's home is owned outright (i.e. without a mortgage). We also control for household wealth, household labor income, household benefit income and the log expected value of the gamble,¹³ as these may influence risk preference, especially if individuals misinterpret 'income' in the above questions M1 to M5 to include wealth. We also include an early measure of risk preference based on actual behaviour, which is defined in detail below. These variables are all contained in the vector \mathbf{Z} and measured at 1996. Thus, we estimate the following equation as an ordered probit regression:

$$r_i = g(\mathbf{Z}_i, \ln y_i) + \varepsilon_{i2} \quad (2)$$

¹³ To be specific: if $RA = 0$, $ev = 0.5(2LY) + 0.5(0.25LY)$; if $RA = 1$, $ev = 0.5(2LY) + 0.5(0.5LY)$; if $RA = 2$, $ev = 0.5(2LY) + 0.5(0.66LY)$; if $RA = 3$, $ev = 0.5(2LY) + 0.5(0.8LY)$; if $RA = 4$, $ev = 0.5(2LY) + 0.5(0.9LY)$; finally, if $RA = 5$, $ev = LY$.

In order to explore the robustness of our findings, we then re-estimate our educational attainment equations replacing r_i with a value purged from identifiable influences, defined as: $\hat{\varepsilon}_{i2}$, the residual from the ordered probit model, i.e. equation (2), where the risk attitudes index is the dependent variable and the explanatory variables represent a combination of individual and household characteristics.

The results from estimating the two risk preference equations (i.e. risk avoidance and risk tolerance) are presented in Tables 5A and 5B. It is apparent from the estimated risk preference equations, i.e. equation (2), that RA_i and RT_i are both influenced by age which suggests that risk preferences are not time invariant. However, the marginal effects of the quadratic in age (see Table 5B) are relatively small for both measures of risk preference at the extreme values of the two measures and are only statistically significant at the 10 per cent level for RT_i and insignificant for RA_i .

We replicate our analysis of Table 3 by estimating equation (1) based upon $\hat{\varepsilon}_{i2}$, the residual from equation (2), for both measures of education. Table 6 Panels A and B present the results for the highest education attainment index (e_i) whilst Panel C presents the results for years of completed schooling (s_i).¹⁴ The results summarized in Table 6 concur with those of Table 3, where risk preference is treated exogenously, in that there is a statistically significant relationship between both measures of risk preference and education. For example, a one standard deviation increase in $\hat{\varepsilon}_{i2}$ is associated with an increase in the probability of having less than high school education by 1%. Similarly, at the opposite end of the educational attainment hierarchy, a one standard deviation increase in $\hat{\varepsilon}_{i2}$ is associated with a decrease in the probability of having completed postgraduate study by 0.52%. Hence, our

¹⁴ Our use of a generated variable may potentially induce bias in the estimates. As such, the standard errors on $\hat{\varepsilon}_{i2}$ in equation (1) have been bootstrapped with 200 replications.

findings suggest that the correlation between risk preference and education reported in Table 3 is not capturing, for example, an unobserved wealth or age effect.

To further explore this issue, we estimate educational attainment equations by age cohorts, i.e. by individuals born in the following decades: 1930s; 1940s; 1950s; 1960s and 1970s. We test whether the influence of risk preference on educational attainment in each cohort (i.e. sub-sample) is significantly different from the estimated coefficient on risk preference in the education equation estimated over the entire sample. The results presented in Table 7 reveal that the null hypothesis that the influence of risk preference on education is the same as the 1996 effect of risk preference on education cannot, in general, be rejected across successive cohorts.¹⁵ The findings that the estimated coefficients on the risk preference measures do not vary across the age cohorts provide further support for the time invariance of the influence of risk preference on educational attainment.

Finally, over the period 1969 to 1972, an index of risk avoidance is available in four waves of the *PSID*, which is the early risk preference measure included in equation (2) above. This measure of risk avoidance is derived from questions relating to factors such as the head of household's seat belt usage, smoking behavior and purchases of medical insurance and car insurance, i.e. actual rather than hypothetical behavior. It is possible that individuals are in the sample between 1 to 4 times during the period 1969 to 1972. Hence, we take an average of the risk avoidance index over a maximum of four years as our early measure of risk preference, RA_i^o . There are 647 individuals who were heads of households in both 1996 and over the period 1969 to 1972. These individuals are aged between 41 and 65. Hence, for these individuals we can compare the relationship between the risk preference measure reported in 1996 and educational attainment with that of the risk preference measure reported over 1969

¹⁵ Tests that the estimated coefficients on the risk preference variable between successive cohorts, 1930=1940; 1940=1950 etc., are equal cannot be rejected.

to 1972.¹⁶ Indeed, if risk preference is largely time invariant then we would expect the relationship between RA_i^o and education to have the same sign and to be similar in magnitude to that between RA_i and education. It should be explicitly acknowledged that RA_i and RA_i^o do differ in terms of the underlying survey questions being based on hypothetical behavior in the case of RA_i and on actual behavior in the case of RA_i^o . Despite such differences, however, for this sub-sample of individuals, the correlation between the risk aversion index of 1996 and risk avoidance index of the earlier period is 0.0797, which is statistically significant at the 5 per cent level. Thus, the two risk preference variables, although constructed from survey responses given two decades apart, are positively related suggesting time invariance of risk preferences.

To further investigate the relationship between risk preference and human capital, we estimate equation (1) based on the sub-sample of 647 individuals who were present in both the 1996 *PSID* and at least one year in the 1969 to 1972 *PSID*. The results are shown in Table 8 for both measures of education: Panels A and B present the results for the highest education attainment index (e_i) for each risk measure, and Panel C presents the results for years of completed schooling (s_i). In Panels A and B, the marginal effects are reported across each of the education categories, along with the percentage impact of a one standard deviation increase in risk preference. For this sub-sample of individuals, risk preference measured at 1996 and risk preference measured over 1969 to 1972 are both statistically significantly related to each category of educational attainment and the number of years of completed schooling. Noticeably, the association between risk preference and educational attainment is more pronounced for the earlier period relative to 1996: 6.49% versus 1.9% for the probability of having less than high school education ($e_i = 0$); 7.19% versus 1% for the

¹⁶ The risk avoidance index is increasing in risk aversion and so *a priori* we would expect it to be positively correlated with the 1996 measure of risk aversion.

probability of having completed some postgraduate education ($e_i = 4$); and 17.08% versus 1.95% for years of completed schooling. Such findings are not surprising as one might predict that the majority of educational attainment would have been achieved closer to the early time period and, hence, the relationship between RA_i^o and human capital accumulation is predictably stronger than that between risk preference measured in 1996 and educational attainment.¹⁷

V. Parental Risk Preference and Children's Academic Achievement

Given that parents play an important role in decisions regarding their children's education, it is apparent that a parent's risk preference may influence their off-spring's education, potentially to a greater extent than the risk preference of the child. To be specific, an individual's educational attainment may reflect the decisions made on his/her behalf by parents and, hence, may reflect the risk preference of the parents. In order to explore this hitherto neglected area of research, we exploit the data from the *PSID Child Development Supplement (CDS) 1997*. The 1997 *CDS* provides additional information relating to parents in the *PSID* and their children with the objective being to provide information on early human capital formation.¹⁸ All *PSID* families with children aged between 0 and 12 were invited to complete the *CDS*, where up to two children per family were included in the survey. In cases where there were more than two eligible children in the family, two were randomly selected to take part in the study. Our sample of children from the 1997 *CDS* comprises approximately 1,000 children. We match the sample of children to the 1996 *PSID*, which provides detailed information on their parents, in particular, the risk preference of their parents.

¹⁷ We also explore this by including the two risk preference measures, the 1996 measure and the 1969 to 1972 measure, in the educational attainment equation simultaneously. The equality of the coefficients is always rejected at the 1 per cent level across both measures of education, with the early measure having the dominant effect.

¹⁸ A number of papers have exploited the detailed information in the *PSID CDS*. For example, Weinberg (2001) has explored an incentive model of the effect of parental income on children reporting a positive relationship between parental income and child outcomes. Case *et al.* (2002), using the *PSID CDS*, show that the relationship between income and health status for adults has antecedents in childhood.

We focus on the relationship between the parent's risk preference and their children's age-standardized scores in the Woodcock-Johnson Revised Achievement Tests, which are widely used and have been validated extensively (see Woodcock and Johnson, 1990, for further details of the tests). As part of the 1997 *CDS*, children aged 3 to 12 took the Woodcock-Johnson Achievement Test, covering: Reading Tests and Mathematics Test. The Reading Test is a combination of a Letter Word Identification Test and a Passage Comprehension Test; similarly, the Mathematics Test is a combination of an Applied Problems Test and a Calculation Skills Tests. Children younger than 6 years old did not complete all the tests, therefore we focus our study on the Standardized Applied Problem Test, with a sample of 1,038 children (mean age 7 years old) and the Standardized Reading Test with a sample of 722 children (mean age 9 years old).

We explore the relationship between parental risk preference and their child's achievements in the Standardized Reading Test and the Standardized Applied Problem Test by modeling the child's (j) 1997 test score, $TEST_j$, as a function of the risk preference of the parent who is the head of household in 1996 (i) employing OLS, as follows:

$$TEST_{j1997} = h(\mathbf{K}_j, r_{i1996}) + \varepsilon_j \quad (3)$$

where r_i denotes the measure of parental risk aversion elicited from the *PSID* 1996, and \mathbf{K}_j represents a set of additional explanatory variables (derived from the 1997 and 2002 *CDS*), which includes information related to the child such as: age; weight; gender; ethnicity; whether the child is living with his/her parents; and the number of children in the household. In this set of explanatory variables, we also include variables, which are related to the parent including: marital status; religion; years of education; household labor income, household wealth, and household income from benefits. The results presented in Table 9, Panel A, suggest that scores in the Reading and Applied Problem Tests are inversely associated with

the parent's risk preference index (r_i), where a one standard deviation increase in RA_i is associated with a around a 6% (9%) lower reading (applied problem) test score. In Panel B, we replace the risk aversion index with the risk tolerance index. The results suggest that risk tolerance is positively associated with the test scores. The estimated relationship between the risk aversion/risk tolerance index and the test scores does not change when we replace the index with its residual ($\hat{\varepsilon}_{i2}$), constructed as described above, as shown in Panels C and D.

Finally, we split the sample according to whether the head of the household is the father or the mother of the child, as potentially this might influence the relationship between parental risk preference and the child's test score. Interestingly, the results presented in Table 10, Panels A and B, suggest that the risk preference of the father has no effect on the child's test scores. However, where the mother is the head of the household, Panels C and D, the inverse relationship between the risk aversion of the parent and the test scores of their offspring is statistically significant.

VI. Conclusions

This paper has focused on the relationship between risk preference and educational attainment using individual level U.S. data drawn from the *PSID*. Our empirical findings support a statistically significant relationship between risk preference and educational attainment. This result is robust across different measures of education and different measures of risk preference. Specifically, greater levels of risk aversion are inversely associated with educational attainment. Our findings are consistent with the theoretical literature and the limited amount of empirical evidence in this area, such as, Barsky *et al.* (1997), and Guiso and Paiella (2007). In addition, we explore the relationship between the risk preference of parents and the educational achievements of their children using the 1997 *PSID CDS*. Despite the important role of the parental decision-making in their off-spring's education, this relationship has not been previously studied in the empirical literature. Our findings support

an inverse relationship between parental risk aversion and the educational achievements of their children.

Given both the potential time and financial dimensions to the investments made by parents in their children's education, our findings are particularly interesting from a policy-maker's perspective. For example, our findings add to the current debate on the funding and access to higher education especially in the context of the reforms to the funding for higher education in the U.K., which have been designed to alter the social mix of students to encourage participation amongst lower socio economics groups, Greenaway and Haynes (2000). If risk-aversion is concentrated amongst the lower socio-economic groups, then our framework predicts that such individuals may be unlikely to invest in their own human capital or the human capital of their children given the current funding system.

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Table 1: Correlation Matrix for Risk Preference and Educational Attainment

	RA_i	RT_i	e_i	s_i
RA_i	1.0000			
RT_i	-0.9533 $p=[0.000]$	1.0000		
e_i	-0.0554 $p=[0.003]$	0.0482 $p=[0.001]$	1.000	
s_i	-0.1062 $p=[0.000]$	0.0838 $p=[0.000]$	0.6326 $p=[0.000]$	1.000

Table 2: Summary Statistics

	<u>MEAN</u>	<u>STD. DEV</u>	<u>MAX</u>	<u>MIN</u>
<i>Head of household characteristics in 1996:</i>				
Highest educational attainment index e_i	2.6192	1.1978	5	1
Number of years of completed schooling s_i	12.7998	3.1909	17	0
Risk aversion index RA_i	3.1760	1.6314	5	0
Risk tolerance RT_i	1.0640	1.2014	3	0
Age	39.3087	10.8307	65	18
Male	0.7218	0.4481	1	0
White	0.3945	0.4887	1	0
Black	0.2209	0.4149	1	0
Latin	0.0077	0.0878	1	0
Catholic	0.1356	0.3424	1	0
Jewish	0.1762	0.1315	1	0
Protestant	0.4508	0.4976	1	0
Log household wealth	2.11872	3.3328	14.5086	0
Log household labor income	9.5288	2.4886	13.5923	0
Log household income from benefits	0.98171	2.5546	13.8155	0
Log expected value of the gamble	13.5725	2.3914	17.5314	0
Own home	0.35798	0.4936	1	0
Number of children in the household	0.9840	1.1515	8	0
Household size	2.7951	1.4243	10	1
Single	0.2014	0.4011	1	0
Separated/divorced	0.2122	0.4089	1	0
Widowed	0.2899	0.1678	1	0
<i>Head of household's childhood:</i>				
Mother single when child born	0.4813	0.2140	1	0
Mother widow when child born	0.0036	0.0599	1	0
Mother separated/divorced when child born	0.1478	0.1206	1	0
Lived with parents until 16	0.6236	0.4845	1	0
Mother worked when child growing up	0.3213	0.4670	1	0
Father worker when child growing up	0.9990	0.0307	1	0
Father professional or managerial	0.0860	0.2804	1	0
Father self employed	0.4737	0.2124	1	0
Father clerical or crafts	0.2078	0.4058	1	0
Father manual	2.1320	3.6895	1	0
Number of siblings	2.1315	2.4935	10	0
Firstborn	0.1589	0.3657	1	0
Mother born outside US	0.2931	0.4552	1	0
Mother high school education	0.3359	0.4723	1	0
Mother college education	0.1303	0.3367	1	0
Father high school education	0.2141	0.4102	1	0
Father college education	0.1627	0.3692	1	0
Family was poor when growing up	0.3215	0.4671	1	0
OBSERVATIONS		5,277		

Table 3: Risk Preference and Educational Attainment

PANEL A: HIGHEST EDUCATION ATTAINMENT INDEX (e_i) AND RISK AVERSION INDEX										
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Risk Aversion Index RA_i	0.0118	(5.22)	0.0073	(5.13)	-0.0052	(5.13)	-0.0076	(5.16)	-0.0062	(5.16)
Chi Squared (28)	934.55 $p=[0.000]$									
Pseudo R Squared	0.0582									
PANEL B: HIGHEST EDUCATION ATTAINMENT (e_i) AND RISK TOLERANCE										
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Risk Tolerance RT_i	-0.0134	(4.39)	-0.0082	(4.33)	0.0595	(4.31)	0.0086	(4.35)	0.0071	(4.35)
Chi Squared (28)	926.47 $p=[0.000]$									
Pseudo R Squared	0.0577									
PANEL C: YEARS OF COMPLETED SCHOOLING INDEX AND RISK AVERSION (s_i)										
	<u>COEF</u>				<u>TSTAT</u>					
Risk Aversion Index RA_i	-0.0748				(2.86)					
Risk Tolerance RT_i	0.0860				(2.44)					
Adjusted R Squared	0.1079				0.1027					
OBSERVATIONS	5,277									

Notes: (i) Control variables are as shown in Table 4; (ii) M.E. denotes marginal effect; (iii) The results shown in Panels A and B are estimated from an ordered probit model, whilst those in Panel C are from OLS estimation.

Table 4: The Determinants of Educational Attainment

	HIGHEST EDUCATION ATTAINMENT INDEX e_i		YEARS OF COMPLETED SCHOOLING INDEX s_i	
	COEF	TSTAT	COEF	TSTAT
<i>Head of household characteristics in 1996:</i>				
Age	0.7088	(8.52)	0.1784	(7.66)
Age squared	-0.0007	(7.68)	-0.0019	(7.08)
Male	0.0912	(2.50)	0.5339	(5.15)
White	-0.0494	(0.93)	0.0552	(0.36)
Black	-0.1791	(2.86)	-0.3123	(1.75)
Latin	-0.4661	(2.58)	-1.1241	(2.24)
Catholic	0.2462	(4.07)	0.5530	(3.20)
Jewish	0.6842	(5.57)	1.2475	(3.60)
Protestant	0.0645	(1.29)	0.1860	(1.31)
Risk Aversion Index	-0.0479	(5.23)	-0.0748	(2.86)
<i>Head of household's childhood:</i>				
Mother single when child born	-0.0859	(1.16)	-0.0799	(0.38)
Mother widow when child born	0.0389	(0.16)	0.1106	(0.16)
Mother separated/divorced when child born	-0.0150	(0.12)	0.0264	(0.08)
Lived with parents until 16	0.0864	(2.71)	0.3409	(3.74)
Mother worked when child growing up	0.0348	(1.07)	0.0908	(0.97)
Father worked when child growing up	-0.4479	(0.92)	-0.7731	(0.57)
Father professional or managerial	0.3798	(6.64)	0.8194	(5.00)
Father self employed	0.3160	(4.45)	0.6861	(3.37)
Father clerical or crafts	0.0156	(0.41)	0.0425	(0.39)
Father manual	-0.0596	(1.45)	-0.1162	(0.99)
Number of siblings	-0.0590	(9.05)	-0.1207	(6.58)
Firstborn	0.1273	(2.92)	0.3080	(2.46)
Mother born outside US	0.6702	(1.88)	0.2327	(2.29)
Mother high school education	0.1383	(3.86)	0.2900	(2.83)
Mother college education	0.3523	(6.76)	0.8335	(5.58)
Father high school education	0.1232	(3.08)	0.2672	(2.33)
Father college education	0.3243	(6.55)	0.5755	(4.05)
Family was poor when growing up	-0.0618	(1.60)	-0.1013	(1.92)
Chi Squared (28)	934.55		–	
Pseudo R Squared	0.0582		–	
Adjusted R Squared	–		0.1031	
OBSERVATIONS		5,277		

Notes: e_i is estimated as an ordered probit model and s_i is estimated by OLS.

Table 5A: The Determinants of Risk Preference

	RISK AVERSION RA_i		RISK TOLERANCE RT_i	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Age	-0.0152	(1.47)	0.0193	(1.71)
Age squared	0.0003	(2.75)	-0.0004	(2.93)
Male	-0.2289	(5.05)	0.2423	(4.98)
White	-0.0815	(1.80)	0.0577	(1.19)
Black	0.0491	(0.93)	-0.0529	(0.93)
Latin	-0.1850	(1.09)	0.1079	(0.59)
Single	-0.1655	(2.95)	0.1939	(3.22)
Separated/divorced	-0.0794	(1.47)	0.1009	(1.74)
Widow	0.0394	(0.37)	0.0456	(0.39)
Number of children in the household	-0.0200	(0.64)	-0.0185	(0.55)
Household size	0.2477	(0.90)	0.0042	(0.14)
Own home	0.0981	(2.71)	-0.1080	(2.77)
Early risk preference measure	0.0636	(4.54)	-0.0697	(4.60)
No information about early risk preference	0.0076	(0.17)	-0.0171	(0.36)
Log household wealth	-0.0119	(2.51)	0.0100	(1.97)
Log household labor income	0.0171	(2.73)	-0.0213	(3.18)
Log household income from benefits	-0.0143	(2.39)	0.0155	(2.42)
Log expected value of hypothetical gamble	-0.0154	(1.54)	0.0067	(0.62)
Chi Squared (18)	281.07 $p=[0.000]$		237.41 $p=[0.000]$	
Pseudo R Squared	0.0157		0.0180	
OBSERVATIONS	5,277			

Table 5B: The Determinants of Risk Preference – Marginal Effects of Age

PANEL A: Risk Aversion Index RA_i	LOWEST ($RA_i = 0$)		HIGHEST ($RA_i = 5$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Age	0.0018	(1.47)	-0.0053	(1.47)
Age Squared	-0.0001	(2.74)	0.0001	(2.75)
PANEL B: Risk Tolerance RT_i	LOWEST ($RT_i = 0.15$)		HIGHEST ($RT_i = 0.57$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Age	-0.0077	(1.71)	0.0053	(1.71)
Age Squared	0.0001	(2.93)	-0.0001	(2.93)

Notes: (i) Control variables in Table 5B are as in Table 5A; (ii) M.E. denotes marginal effect.

Table 6: Risk Preference Residual ($\hat{\varepsilon}_{i2}$) and Educational Attainment

PANEL A: HIGHEST EDUCATION ATTAINMENT INDEX (e_i) AND RISK AVERSION												
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)			
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>		
Risk Aversion Index – Residual $\hat{\varepsilon}_{i2}$	0.0098	(4.32)	0.0060	(4.26)	-0.0043	(4.24)	-0.0063	(4.28)	-0.0052	(4.29)		
Chi Squared (28)	925.86 $p=[0.000]$											
Pseudo R Squared	0.0577											
PANEL B: HIGHEST EDUCATION ATTAINMENT INDEX (e_i) AND RISK TOLERANCE												
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)			
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>		
Risk Tolerance – Residual $\hat{\varepsilon}_{i2}$	0.0540	(2.95)	0.0331	(2.93)	-0.0238	(2.93)	-0.0346	(2.94)	-0.0286	(2.94)		
Chi Squared (28)	915.87 $p=[0.000]$											
Pseudo R Squared	0.0571											
PANEL C: YEARS OF COMPLETED SCHOOLING INDEX (s_i)												
	<u>COEF</u>				<u>TSTAT</u>		<u>COEF</u>				<u>TSTAT</u>	
Risk Aversion Index – Residual $\hat{\varepsilon}_{i2}$	-0.0584				(2.23)							
Risk Tolerance – Residual $\hat{\varepsilon}_{i2}$							-0.5423				(2.57)	
Adjusted R Squared	0.1026						0.1028					
OBSERVATIONS	5,277											

Notes: (i) Control variables are as shown in Table 4; (ii) M.E. denotes marginal effect; (iii) The results shown in Panels A and B are estimated from an ordered probit model, those in Panel C are from OLS estimation.

Table 7: The Relationship between Risk Preference and Education Attainment across Age Cohorts

COEFFICIENT		NULL HYPOTHESIS		DECADE OF BIRTH				
				<u>1930</u>	<u>1940</u>	<u>1950</u>	<u>1960</u>	<u>1970</u>
Dependent Variable = e_i , Risk Preference = RA_i				-0.1434	-0.0958	-0.0248	-0.0166	-0.0687
	= 1996 effect	-0.0445		$p=[0.0651]$	$p=[0.0592]$	$p=[0.2703]$	$p=[0.1295]$	$p=0.3807]$
Dependent Variable = e_i , Risk Preference = RT_i				1.7016	0.8813	0.1790	0.1021	0.7039
	= 1996 effect	0.3894		$p=[0.0284]$	$p=[0.0792]$	$p=[0.2420]$	$p=[0.1201]$	$p=0.2364]$
Dependent Variable = s_i , Risk Preference = RA_i				-0.3436	-0.3120	-0.1121	-0.1123	-0.1328
	= 1996 effect	-0.1285		$p=[0.1113]$	$p=0.0750]$	$p=[0.6648]$	$p=[0.8585]$	$p=[0.9324]$
Dependent Variable = s_i , Risk Preference = RT_i				3.5549	2.7086	0.7412	0.9730	1.2108
	= 1996 effect	1.0096		$p=[0.1145]$	$p=[0.0104]$	$p=[0.4847]$	$p=[0.9172]$	$p=[0.6829]$
OBSERVATIONS				247	861	1,772	1,642	755
(% SAMPLE)				(4.69%)	(16.32%)	(33.58%)	(31.11%)	(14.30%)

Notes: (i) Controls are as in Table 4; (ii) e_i is estimated as an ordered probit model and s_i is estimated by OLS.

Table 8: The Timing of the Measurement of Risk Preference and Educational Attainment

PANEL A: HIGHEST EDUCATION ATTAINMENT INDEX (e_i) AND RISK AVERSION (1996)										
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Risk Aversion Index RA_i	0.01949	(3.46)	0.01409	(3.33)	0.00691	(2.92)	-0.01881	(3.35)	-0.02167	(3.48)
Effect (%)	2.95%		2.13%		1.04%		-2.84%		-3.28%	
Chi Squared (35)	120.62 $p=[0.000]$									
Pseudo R Squared	0.0601									
PANEL B: HIGHEST EDUCATION ATTAINMENT INDEX (e_i) AND RISK AVOIDANCE INDEX (1969-72)										
	LESS THAN HIGH SCHOOL COMPLETED ($e_i = 0$)		HIGH SCHOOL COMPLETED ($e_i = 1$)		WENT TO COLLEGE DID NOT GRADUATE ($e_i = 2$)		GRADUATED FROM COLLEGE ($e_i = 3$)		SOME POSTGRADUATE STUDY ($e_i = 4$)	
	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>	<u>M.E.</u>	<u>TSTAT</u>
Risk Avoidance Index RA_i^o	0.04662	(6.75)	0.03638	(6.14)	0.01888	(4.41)	-0.05021	(6.21)	-0.05167	(6.95)
Effect (%)	6.49%		5.06%		2.63%		-6.99%		-7.19%	
Chi Squared (35)	166.40 $p=[0.000]$									
Pseudo R Squared	0.0827									
PANEL C: YEARS OF COMPLETED SCHOOLING INDEX (s_i)										
	<u>COEF</u>	<u>TSTAT</u>	<u>Effect (%)</u>	<u>COEF</u>	<u>TSTAT</u>	<u>Effect (%)</u>				
Risk Aversion Index RA_i	-0.25661	(4.04)	6.95%	-0.72379	(10.67)	17.08%				
Risk Avoidance Index RA_i^o										
OBSERVATIONS	647									

Notes: (i) Control variables are as shown in Table 4; (ii) M.E. denotes marginal effect relating to the probability of having no education, i.e. less than high school; (iii) The results shown in Panels A and B are estimated from an ordered probit model, those in Panel C are from OLS estimation; (iv) in Panels A and B the "Effect (%)" figures are derived by multiplying the marginal effect by the standard deviation of the relevant risk preference measure; (v) in Panel C the figures in the column labeled 'Effect (%)' are derived from $(\partial s / \partial r) \times (\bar{r} / \bar{s})$, i.e. risk preference elasticity, where $\hat{\phi} = (\partial s / \partial r)$ and \bar{r} and \bar{s} denote the mean values of r and s respectively.

Table 9: Children's Academic Test Scores and Parent's Risk Aversion**PANEL A: DETERMINANTS OF TEST SCORES AND PARENT'S RISK AVERSION**

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Age	0.2693	(4.30)	0.4288	(2.43)
Age squared	-0.0150	(3.64)	-0.0225	(2.30)
Weight	0.0079	(3.10)	0.0045	(1.47)
Male	-0.0716	(1.19)	0.0295	(0.40)
White	0.2838	(3.48)	0.2437	(2.39)
Living with parents	-0.0852	(0.69)	0.1015	(0.60)
Number of children in the household	-0.0236	(0.74)	-0.0014	(0.04)
Parent: Single	-0.2293	(1.55)	0.1046	(0.55)
Parent: Separated/divorced	0.0991	(0.68)	0.3441	(1.77)
Parent: Widow	0.0252	(0.08)	0.0760	(0.19)
Parent: Catholic	0.1246	(2.38)	0.2259	(3.45)
Parent: Jewish	-0.0609	(0.28)	-0.0562	(0.22)
Parent: Protestant	-0.0271	(0.41)	0.0064	(0.08)
Years of education of the head of household	-0.0311	(2.71)	-0.0020	(0.14)
Log household labor income	-0.0144	(1.12)	-0.0094	(0.60)
Log household wealth	0.0043	(0.46)	0.0040	(0.35)
Log household income from benefits	-0.0200	(1.65)	-0.0124	(1.76)
Parent's risk aversion index RA_i	-0.0322	(1.87)	-0.0485	(2.00)
Adjusted R Squared	0.0715		0.0354	

PANEL B: DETERMINANTS OF TEST SCORES AND PARENT'S RISK TOLERANCE

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Parent's risk tolerance index RT_i	0.0521	(1.99)	0.0600	(1.88)
Adjusted R Squared	0.0727		0.0347	

PANEL C: DETERMINANTS OF TEST SCORES AND PARENT'S RISK AVERSION-RESIDUAL

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Parent's risk aversion index – Residual $\hat{\varepsilon}_{i2}$	-0.0330	(1.96)	-0.0486	(2.00)
Adjusted R Squared	0.0716		0.0353	

PANEL D: DETERMINANTS OF TEST SCORES AND PARENT'S RISK TOLERANCE -RESIDUAL

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Parent's risk aversion tolerance – Residual $\hat{\varepsilon}_{i2}$	0.0258	(1.92)	0.0255	(1.81)
Adjusted R Squared	0.0716		0.0322	
OBSERVATIONS	1,038		722	

Notes: Panels B-D contain the same control variables as given in Panel A.

Table 10: Children's Academic Test Scores and Parent's Risk Aversion**PANEL A: DETERMINANTS OF TEST SCORES AND FATHER'S RISK AVERSION**

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Father's risk aversion index RA_i	-0.0108	(0.38)	-0.0347	(1.03)
Adjusted R Squared	0.0719		0.0467	

PANEL B: DETERMINANTS OF TEST SCORES AND FATHER'S RISK TOLERANCE

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Father's risk tolerance index RT_i	0.0234	(0.64)	0.0401	(0.91)
Adjusted R Squared	0.0723		0.0461	
OBSERVATIONS	527		378	

PANEL C: DETERMINANTS OF TEST SCORES AND MOTHER'S RISK AVERSION

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Mother's risk aversion index RA_i	-0.0623	(2.22)	-0.0648	(1.82)
Adjusted R Squared	0.0784		0.0053	

PANEL D: DETERMINANTS OF TEST SCORES AND MOTHER'S RISK TOLERANCE

	APPLIED PROBLEMS		READING	
	<u>COEF</u>	<u>TSTAT</u>	<u>COEF</u>	<u>TSTAT</u>
Mother's risk tolerance index RT_i	0.0942	(2.51)	0.0836	(1.76)
Adjusted R Squared	0.0809		0.0046	
OBSERVATIONS	511		344	

Notes: Panels A-D contain the same control variables as given in Table 9 Panel A.