# WERE MANKIW, ROMER, AND WEIL RIGHT? A RECONCILIATION OF THE MICRO AND MACRO EFFECTS OF SCHOOLING ON INCOME

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In Mankiw, Romer, and Weil's augmented Solow model [Quarterly Journal of Economics 107 (2) 407–437 (1992)], the marginal product of human capital accrues to three factors of production: directly to human capital, and as an external effect to physical capital and labor. This paper estimates national stocks of human capital in 1990 created from prior investment in schooling and shows that for 36 countries the (macro) marginal product of human capital accruing to workers in 1990 is consistent with estimates of the (micro) marginal return on investment in schooling in workers' earnings studies. This reconciliation provides empirical evidence for the augmented Solow model.

Keywords: Human Capital, Education, Neoclassical Model, Economic Growth

### 1. INTRODUCTION

When Mankiw, Romer, and Weil (1992) added human capital from schooling to the Solow model, they solved several empirical problems. This addition enabled the model to explain differences in income across countries, to provide an accurate estimate of the share of national income accruing to physical capital, and to provide a more accurate representation of the rate of income convergence in response to changes in rates of capital investment. But it also revealed a statistical relationship that many found problematic—a very large effect of human capital on national income.

In Mankiw, Romer, and Weil's version of the Solow model (hereafter denoted the MRW model), productivity (*A*) varies over time, but country-specific differences affecting productivity not related to physical or human capital levels are assumed to be random:<sup>1</sup>

$$(Y/L)_{it} = (K/L)_{it}^{\alpha} (H/L)_{it}^{\beta} (A_t)^{1-\alpha-\beta}.$$
 (1)

MRW's empirical results, since replicated by other researchers,  $^2$  indicate that differences in physical capital (K), human capital (H), and labor (L) can explain almost all of the variation in national income (Y) across countries, leaving little

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or no variation to be explained directly by other factors. In this model the levels of physical capital, human capital, and labor are the proximate determinants of national income. Other national characteristics, including institutions, policies, and culture, affect income through their effect on these factors.<sup>3</sup>

Critics have argued that the MRW model is mis-specified because it does not include other variables, but the evidence supporting the inclusion of other variables is weak. Levine and Renelt (1992) examined the relationship between over 50 variables and growth and found that over the period 1960–1989, only investment in physical capital was robustly correlated with growth. Ciccone and Jarocinski (2010) evaluated 67 variables and found that over the period 1960–1996, only education variables are robustly linked with growth.

Klenow and Rodriguez-Clare (1997) and Hall and Jones (1999) contributed to the perception that variables are omitted from the model. They presented analyses that purportedly show that the effect of schooling on workers' earnings in micro studies is much smaller than the effect in MRW's empirical results. They estimated the average effect of schooling on earnings in micro studies and then subtracted these earnings and their estimate of the effect of physical capital from national income to estimate country-specific productivity residuals  $(A_i)$ . Klenow and Rodriguez-Clare used the MRW model to estimate this residual, whereas Hall and Jones used a slightly different model, but both analyses show that national differences in productivity  $(A_i)$  have a larger effect on national income than differences in workers' earnings related to schooling.

Both sets of authors demonstrate that their productivity residuals are highly correlated with human capital. Klenow and Rodriguez-Clare acknowledge that these residuals could be an external effect of human capital related to the speed of technology diffusion, but they show that for the MRW empirical results to be valid, this effect would have to be much larger than the direct effect on workers' earnings. Hall and Jones argue that the productivity residuals are caused by differences in institutions and government policies, and they present statistical evidence to support this view. These papers were influential in undermining support for MRW's model, which is unfortunate, because the methodology used to calculate the productivity residuals is flawed.

The MRW model presumes that physical capital and skill are complementary. An increase in the level of one capital factor increases the marginal product of the other capital factor. In any model with a multiplicative relationship between the two forms of capital, the external effects of human capital (and physical capital) are an integral part of the model, and any comparison of the effect of schooling in micro studies with the empirical results in these models must account for these external effects. Because of the complementary effect of skill on (physical) capital, only part of the marginal product of human capital accrues to workers. Because they did not take this into account, Klenow and Rodriguez-Clare and Hall and Jones substantially overestimated the productivity residuals  $(A_i)$  remaining after the micro estimates of schooling and physical capital were subtracted from national income.

As shown later in this paper, in MRW's model the aggregate external effect of human capital is  $(1-\beta)/\beta$  times the direct effect. Given the independent estimate that  $\alpha$  is approximately 0.35, the Cobb–Douglas structure with constant returns ensures that  $\beta \leq 0.65$ . As a result, the aggregate external effect of human capital in the model is at least 0.54 times the direct effect (0.35/0.65) and possibly much higher. With MRW's estimate of  $\beta = 0.3$ , the external effect of human capital in the MRW model is 2.3 times the direct effect.

The large external effect of human capital inherent in the MRW model (and in Hall and Jones's model) is either a great strength or a great weakness. But it is not an optional effect, and it cannot be ignored in examining whether MRW's empirical results are consistent with micro estimates of the effect of schooling on workers' income.

In the MRW model human capital (H) has two external effects, one on physical capital (K) and one on labor (L), both of which are mathematically specified. This defined relationship suggests a test of the model's validity. Because workers receive income from human capital through the direct effect  $(\beta)$  and the external effect  $(1-\alpha-\beta)$  on labor, their total income must be bounded by the direct effect on the low side and by the sum of these two effects on the high side. If the MRW model is valid, the micro estimates of workers' incomes must fall within these bounds.

This paper presents the results of such a test. It compares the estimates of the upper and lower bounds of the share of the marginal product of human capital accruing to workers in the MRW model with estimates of the marginal product of schooling in existing micro studies in 36 countries. Conceptually, Klenow and Rodriguez-Clare (1997) and Hall and Jones (1999) performed a similar test, but their implementation of this test was flawed.

If the marginal product of human capital in the macro model is consistent with the marginal product of schooling in the micro studies, then MRW's model, including its large external effects of human capital, is not rejected. If MRW's model is valid, then institutions and policies affect national income indirectly through their effect on the levels of capital, and they are not omitted variables in the model.

This test of the MRW model is a stringent one. The Cobb-Douglas structure with two capital variables is quite constraining. Conceptually the exponents on the two capital stocks are both (1) the marginal effect on national income of an increase in the stock of each type of capital and (2) the share of any increase in national income that accrues either directly or indirectly to that type of capital as a result of an increase in any factor. If the MRW model can explain the distribution of national income across countries and the share of this income that has accrued to each factor of production in these countries, this is a significant achievement.

To perform this test, I estimate the marginal product of human capital in the MRW model in 1990 with a new set of human capital data that are based on each country's cumulative investment in schooling, adjusted for purchasing power parity. My financial measure of human capital is conceptually identical to standard

measures of the stock of physical capital. This measure eliminates many of the estimation issues associated with proxies for human capital, such as years of schooling, which may not be a consistent measure across countries. I include foregone student earnings in both the human capital and the national income data. Subsequently, I estimate the marginal product of schooling in micro studies of workers' earnings from the period 1986–1994 in 36 countries and compare these estimates with the relevant shares of the marginal product of human capital in the MRW model in the same 36 countries.

In contrast to Klenow and Rodriguez-Clare (1997) and Hall and Jones (1999), I find that the estimates of the effect of schooling in MRW's model are consistent with the estimates of the effect of schooling in micro studies. In the lowest-income countries, the marginal product in micro studies is equal to the direct share of the marginal product of human capital (i.e., the share  $\beta$  that accrues to the human capital factor). In the highest-income countries, the marginal product in micro studies is equal to this direct share plus most of the external share that accrues to labor (i.e., the share  $1 - \alpha$ ). These empirical results indicate that as a country's level of human capital rises, a growing share of the external effect of schooling accrues to the educated portion of the work force.

This paper makes several contributions to the literature. First, it provides a conceptual framework for comparing the macro and the micro effects of schooling on income, two strands of the economics literature that previously have not been reconciled. Second, it provides estimates of the stock of human capital from schooling and the marginal product of human capital from schooling for 61 countries in 1990. Third, it shows that the marginal product of human capital in the augmented Solow growth model is consistent with the marginal product of schooling in workers' earnings' studies.

The paper is organized as follows: Section 2 describes the methodology used to create the two sets of rates. Section 3 develops estimates of the stock of human capital from schooling. Section 4 develops new estimates of  $\alpha$  and  $\beta$ . Section 5 develops national average marginal products of schooling from rates of return on investment in schooling in workers' earnings studies. Section 6 compares the micro and macro marginal products of schooling for these countries. Section 7 concludes.

### 2. METHODOLOGICAL CONSIDERATIONS

The rates of return in workers' earnings studies and the marginal product in the Cobb—Douglas national income model are similar in that they both are marginal rates. Both of them estimate the effect of an increase in schooling on income in one year and convert this effect into a marginal product. Despite these similarities, there are differences between these rates that must be addressed to make them comparable:

(1) The macro rates include the national income that accrues to workers and to physical capital, whereas the micro rates include only the part that accrues to workers.

- (2) The average experience of the workers is not the same in the two analyses. The micro rates assume a level of experience that is the average over a worker's life. The macro rates implicitly are based on a lower average level of experience because of the historic growth of population across countries.
- (3) The micro rates implicitly or explicitly account for students' foregone earnings, whereas data on national income exclude these earnings.
- (4) The micro financial rates of return are estimated for schooling at the primary, the secondary, and the university levels, whereas the macro rates estimate the marginal product on aggregate investment in schooling at all levels.
- (5) Some of the micro rates are estimated using the Mincerian methodology, which estimates wage effects from increases in years of schooling rather than financial rates of return [Psacharopoulos and Patrinos (2004)].

In this paper I adjust the input data or the macro and micro estimates of the marginal product of schooling to account for these differences, so that the final estimates are comparable from a conceptual standpoint. To further ensure comparability I develop the macro and micro estimates for the same period of time. The marginal product of schooling in the workers' studies is estimated for 36 countries for which data were available for the period 1986–1994. The relevant shares of the marginal product of human capital in the macro model are estimated for the same countries using data for 1990, the midpoint in the period used for the micro rates. The use of a financial measure for human capital in the macro analysis facilitates the reconciliation of the macro marginal product with the financial rates of return in the micro studies.

### 2.1. Macro Rates of Return

The principal conceptual issue in the analysis is the identification of the share of the marginal product of schooling in the MRW model that is comparable to the rates of return on investment in workers' earnings studies. This share can be identified because the Cobb—Douglas structure in equation (1) provides an explicit distribution of the national income from the marginal product of human capital to the three factors of production.

The marginal product of human capital in the MRW model is

$$MPH_i = \delta Y_i / \delta H_i = \beta Y_i / H_i.$$
 (2)

As is shown subsequently, in the Cobb–Douglas production function, the share of the marginal product of schooling that accrues to each factor of production is equal to the exponent on that factor.

As the marginal product is a partial derivative, it holds the stocks of physical capital and labor constant. As a result, the increase in national income that accrues to physical capital and labor is provided through an increase in the return  $(\Delta r_k)$  on physical capital<sup>4</sup> and an increase in the wage  $(\Delta w)$  paid to labor. The increase in income that accrues to human capital from schooling is the net result of the

increase in the stock of human capital and the decline in the return  $(r_h)$  paid to this stock of capital because of diminishing returns  $(\Delta(r_h H))$ .

From an economic standpoint, an increase in human capital raises the return to physical capital and the wage paid to labor by increasing the productivity of these two factors through an external or spillover effect. In the Cobb-Douglas model, the wage paid to labor in a market economy is equal to the marginal product of labor:

$$w_i = \delta Y_i / \delta L_i = (1 - \alpha - \beta) Y_i / L_i. \tag{3}$$

Mathematically, this wage is the share of national income  $(1 - \alpha - \beta)$  that is left over after the two forms of capital receive their marginal returns. The wage (w) increases when either stock of capital increases because the marginal products of physical capital and human capital in the Cobb-Douglas model are less than the average products.

The wage (w) in the model is the average payment to the labor force resulting from the external effects of physical capital and human capital that does not accrue to those two factors. This wage (w) is separate from the payment made to workers to compensate them for their higher productivity related to their own level of schooling  $(r_h)$ . In the model in equation (1), schooled workers receive both a labor wage (w) and a direct return on the nation's investment in their schooling  $(r_h)$ .

The Cobb–Douglas structure does not specify how the payments to workers from the external effects (w) are distributed to workers with different levels of schooling (including workers with no schooling), nor does it require that these payments be distributed in the same way across countries or within countries over time. If the change in the wage  $(\Delta w)$  caused by the external effect of schooling is provided to all workers (L), then the incremental earnings that schooled workers receive from incremental schooling will include only the direct return on investment in schooling  $(r_h)$ . In this case the rate of return on schooling in workers' earnings studies will be limited to the share of the marginal product of schooling in the macro model that accrues to the recipients of schooling. This share can be calculated as the partial derivative of the national income paid to the human capital factor of production:

$$Direct_i = \delta(r_{hi}H_i)/\delta H_i = \delta((\beta Y_i/H_i)H_i)/\delta H_i = \beta MPH_i = \beta^2 Y_i/H_i.$$
 (4)

The share of the marginal product that accrues to labor as an external effect of schooling can be calculated as the partial derivative of the national income paid to the labor factor of production (wL):

Ext MPH<sub>i</sub> to 
$$L_i = \delta(w_i L_i)/\delta H_i = \delta((1 - \alpha - \beta Y_i/L_i)L_i)/\delta H_i$$
  
=  $(1 - \alpha - \beta)$ MPH<sub>i</sub> =  $(1 - \alpha - \beta)\beta Y_i/H_i$ . (5)

If the change in the wage  $(\Delta w)$  caused by the external effect of schooling is provided entirely to workers who are schooled, then schooled workers' earnings

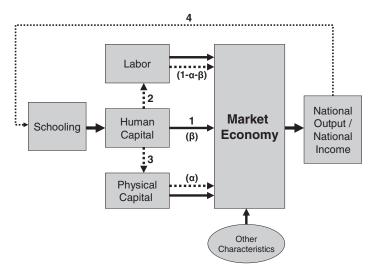


FIGURE 1. Effect of schooling on income in the Mankiw, Romer, and Weil model.

from incremental schooling will include *both* the direct return and the external effect on *w* from investment in schooling. In this case the measured return on investment in schooling in workers' earnings studies will include the direct and the external shares of income that accrue to workers:<sup>5</sup>

Direct<sub>i</sub> + Ext MPH<sub>i</sub> to 
$$L_i = (\beta \text{ MPH}_i) + (1 - \alpha - \beta) \text{MPH}_i$$
  
=  $(1 - \alpha) \text{MPH}_i = (1 - \alpha) \beta Y_i / H_i$ . (6)

So equations (4) and (6) bound the rates of return that should be expected in workers' earnings studies if the MRW model is a valid model and if  $\alpha$  and  $\beta$  are correctly estimated.

Figure 1 provides a schematic of how increased schooling affects national income in the MRW model. The direct effect (1) is shown with a solid line, whereas the external or indirect effects (2, 3, and 4) are shown with a dotted line. In this model the relative magnitudes of the direct effect of schooling on national income and the two external effects (2 and 3) are determined by the estimates of  $\alpha$  and  $\beta$ . The external effects in the Cobb–Douglas model are  $(1-\beta)/\beta$  times the direct effect, so unless  $\beta > 0.5$ , the external effects are larger than the direct effect.

### 2.2. Micro Rates of Return on Investment in Schooling

The rates of return on schooling estimated in workers' earnings studies are typically of two types. One type, which Psacharopoulos and Patrinos (2004) denote as the "full return" rate, is calculated using the standard internal rate of return

methodology used in project analysis. This rate is the discount rate that equalizes the present discounted value of the cost stream and the income stream associated with incremental investment in schooling. The costs include the expenditures on schooling and the students' foregone earnings while they are in school. The income is the incremental earnings associated with the incremental schooling over the life of the worker.

But the full return rates estimated in the literature are not the actual return on investment in schooling, which would be calculated over the life of the investment. These rates are actually the *marginal product* of a certain level of schooling, where income is defined as the average earnings of workers with all levels of experience in one year. This marginal product is conceptually similar to the marginal product in the MRW model. I estimate the marginal product of all schooling in each country by weighting the rate at each level of schooling by the share of national investment that was invested at each level in 1990.

The other rate of return typically found in workers' earnings studies is the "Mincerian" rate, which is an estimate of the relationship between the log of workers' earnings and workers' level of schooling attainment, controlling for work experience, also in one year. Conceptually the Mincerian rate is actually a wage effect, not a return on investment [Psacharopoulos and Patrinos (2004)]. Calculation of the Mincerian rate requires less information than calculation of the full return rate, so Mincerian rates are available for more countries than are full return rates.

The magnitude and the pattern of diminishing returns are very similar for the Mincerian rates and for the marginal product of workers' schooling calculated from the full return rates, so workers' marginal product of schooling can be estimated from the Mincerian rate. In this study there are 16 countries for which there are both full return rates and Mincerian rates. I determine the average ratio between the marginal products and the Mincerian rates and then use this ratio in countries that had only a Mincerian rate to estimate their marginal product of schooling.<sup>6</sup>

### 3. ESTIMATION OF THE NET HUMAN CAPITAL STOCK

Calculation of the MPH in the MRW model requires cross-country data on the stock of human capital. I estimate these stocks using the perpetual inventory method specified in OECD (2001). As far as I know, estimates of a nation's stock of human capital from its investment in schooling have not previously been developed.

The OECD argues that for financial valuation, the appropriate stock of capital is the net capital stock, which includes all of the capital equipment in operation, with a value for this capital that either is a market value, or is calculated using its initial cost (the gross capital stock) and financial depreciation. I calculate each country's gross human capital stock using its aggregate investment in formal schooling, which includes all expenditures related to schooling, financial carrying costs until

the student enters the work force, and students' foregone earnings. I estimate these stocks in 1990 for 61 countries that had been market economies and for which data on school expenditures are available for the period 1950–1985.

In my estimates I assume that students complete their schooling and then work for 40 years. This working life is consistent with the life expectancy of children beginning school in all countries, except in sub-Saharan African countries that have a high incidence of AIDS or other diseases.<sup>7</sup>

I assume that a country's gross human capital stock from schooling in year t is equal to its cumulative national investment in schooling over the years t-45 to t-5. The five-year lag is a rough approximation of the delay from a present discounted value standpoint between the time when a nation invests in schooling and the time when a student enters the work force. This lag is less than half the average student's period of schooling because higher levels of schooling have much higher unit costs than lower levels of schooling [Breton (2010)]. I estimate the cumulative investment using data on public expenditures at five-year intervals as a share of national income, a national ratio of total to public expenditures to account for private expenditures, a common 1.9 ratio of total investment to total expenditures to account for foregone earnings, a national ratio of financial carrying costs for schooling investments, and data on national income:  $^8$ 

$$Gross Hs_{it} = 1.9 * (TotExp/PubExp)_i * 5$$

\* 
$$\sum_{i=1}^{8} [(\text{PubExp}/Y)_{it-5j*} Y_{it-5j*} \text{FinCost}_{it-5j}].$$
 (7)

The data for the ratio of public expenditures to GDP at five-year intervals and the ratio of total to public expenditures for the period 1950-1985 are from Breton (2010), who obtained or estimated these ratios from UNESCO data. The lag between the time when the investment in schooling is made and the time when a student enters the work force has a financial cost, which for physical capital is typically denoted "interest during construction." I include this financial cost in the estimate of each country's gross human capital stock. It is calculated using an annual cost of capital of 8.0% and each country's distribution of schooling attainment in 1990 from Cohen and Soto's (2007) data. The cost of capital is based on Caselli and Feyrer's (2007) finding that the marginal product of reproducible physical capital is similar across countries, with an average of 6.9% in low-income countries and 8.4% in high-income countries. This financial cost factor increases the gross capital investment relative to outlays by multiples that range from 1.3 to 1.6 across countries, depending on the average length of time that students spend in school. These ratios are shown in Appendix B. The derivation of the 1.9 ratio on total expenditures to account for foregone earnings is provided in Appendix A. This ratio is based on 11 studies in nine countries. National income, adjusted for purchasing power parity, is obtained for 1950–1985 from the Penn World Table (PWT) 6.2 data [Heston et al. (2006)].

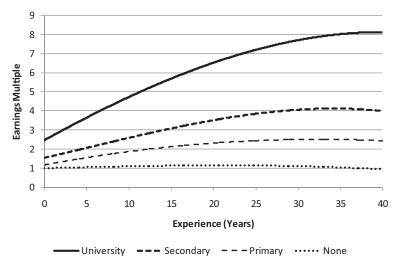


FIGURE 2. Workers' earnings vs. experience by level of schooling.

I estimate net human capital stocks by applying a financial depreciation rate to the gross human capital stocks. The OECD (2001) recommends that financial depreciation be estimated based on the market value of the capital stock over the capital's useful life. Because human capital does not have an observed market value, I estimate the implicit value using the present discounted value of human capital's remaining expected lifetime income.

I assume that human capital's contribution to national income is proportional to its contribution to workers' earnings, and I use worker earnings patterns by level of schooling for a representative set of countries to determine the appropriate depreciation rate. Figure 2 shows the average of the earnings patterns for workers with four levels of schooling in Ecuador, Paraguay, Uruguay, and the Philippines. The present discounted value of the increase in workers' earnings due to schooling indicates that the value of human capital declines linearly over a worker's life, which for a 40-year working life yields an annual depreciation rate of 2.5%. The net human capital stock data created with this depreciation rate are shown in Figure 3 and are provided in Appendix B. The second representation is provided in Appendix B.

The relationship between  $\log(H/L)$  with these new data and average schooling attainment is shown in Appendix D. Although these two measures of human capital are highly correlated,  $\log(H/L)$  is likely to be more accurate than average years of schooling because it accounts for differences in the quality of schooling that are related to differences in investment per year of schooling. Lee and Barro (2001) present evidence that differences in expenditures are correlated with the quality of schooling across countries.

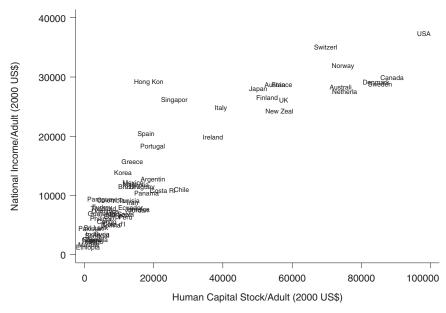


FIGURE 3. National income vs. human capital from schooling in 1990.

# 4. ESTIMATION OF THE MACRO MARGINAL PRODUCT OF HUMAN CAPITAL

I estimate MRW's income model in log form as a model of income per worker to reduce potential heterogeneity bias:

$$\log(Y/L)_i = c + \alpha \log(K/L)_i + \beta \log(H/L)_i + \gamma SSA_i + \varepsilon_i.$$
 (8)

I include a dummy variable for the sub-Saharan African countries (SSA) to control for the effect of their high morbidity and mortality rates, which reduce the effective stock of human capital in these countries. The estimated coefficient on SSA is expected to be negative.

I use estimates per adult for Y/L, H/L, and K/L because all adults contribute directly or indirectly to national income. I use the cumulative investment in physical capital during 1975–1989 (the prior 15 years) as a proxy for the physical capital stock in each country. Because this proxy is highly correlated with the net physical capital stock and the income model is estimated in log form, its use does not bias the results if it is proportional to the actual stock. I estimate national income per adult and physical capital per adult from PWT 6.2 data. <sup>12</sup> I then add foregone student earnings to the income data.

Even though the human capital stock is a predetermined variable, there could be reverse causality between the stock of human capital and national income. Identifying an appropriate instrument for human capital is difficult. Lagged values of human capital are unavailable because of the limited historic data on investment,

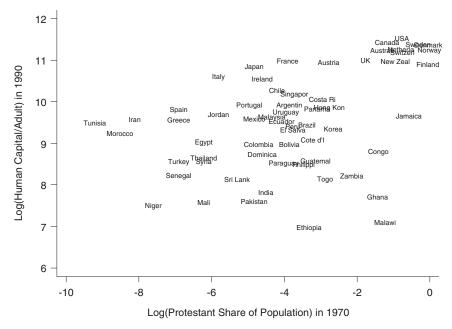


FIGURE 4. Human capital/adult vs. Protestant share of the population.

and they are unlikely to be valid instruments given the lagged response of income to investments in schooling.

I use the log of the Protestant share of the population twenty years earlier (1970) as an instrument for human capital. I obtained the Protestant share data from Barrett (1982). The data for  $\log(H/L)$  and  $\log(\text{Protestant share})$  are shown in Figure 4. The correlation coefficient between these variables is .42. Protestant affiliation is an attractive instrument for human capital in some respects. Most countries have some level of affiliation, and the level of affiliation has been correlated with levels of literacy and schooling across and within countries for centuries [Cipolla (1969), Johansson (1981), Soysal and Strang (1989), Goldin and Katz (1998)]. Given the stability of Protestant affiliation over time, it is not an endogenous variable in an income model.

The disadvantage of Protestant affiliation as an instrument is that it could affect a nation's income through mechanisms not related to investment in schooling. As is well known, Weber (2000) argued that the "Protestant ethic" causes Protestants to save more of their incomes and work harder. What is not well known is that Weber's hypothesis has been overwhelmingly rejected by the empirical evidence [Iannaccone (1998)]. 14

The regression results for the MRW model are shown in Table 1. Column (1) shows the results for an OLS regression. The estimate of  $\alpha$  in this regression (0.38) is consistent with Bernanke and Gurkaynak's (2002) estimate that physical

	(1)	(2)	(3)	(4)	(5)	(6)
Technique	OLS	OLS	2SLS	OLS	OLS	2SLS
Countries	61	61	61	61	57	57
Log(K/L)	0.38*	0.39*	0.34	0.34	0.38*	0.32
	(0.06)	(0.06)	(0.14)	(0.14)	(0.07)	(0.16)
Log(H/L)	0.31*	0.30*	0.36	$0.30^{*}$	0.31*	0.39
	(0.06)	(0.06)	(0.15)	(0.06)	(0.06)	(0.18)
Est $Log(H/L)$				0.06		
				(0.16)		
Sub-Saharan Africa	-0.28*	-0.29*	-0.29*	-0.29*	-0.28*	-0.28*
	(0.11)	(0.11)	(0.11)	(0.11)	(0.11)	(0.11)
Log(Protestant share)		0.01				
		(0.02)				
$R^2$	.95	.95	.95	.95	.95	.94

TABLE 1. Effect of human capital from schooling on national income

*Note*: Dependent variable is Log(national income/adult). White-adjusted standard errors in parentheses. \*Significant at 1% level.

capital's share of national income is about 35% across countries. The coefficients on the capital variables are highly statistically significant, and the model accounts for 95% of the variation in log(Y/L) across the 61 countries.

Column (2) presents the OLS estimates with the addition of the Protestant share variable to investigate whether Protestant affiliation affects national income directly. The direct effect of the Protestant share on national income is negligible, the estimated coefficients on the physical capital and human capital variables are virtually unchanged, and the variation in national income explained by the model is unchanged. There is no indication that Protestant affiliation affects national income other than through its effect on investment in schooling.

Column (3) presents the 2SLS estimate of the model. The coefficients on physical capital and human capital are similar to the OLS estimate, although the estimate of  $\beta$  is larger. The Hausman test in Column (4) indicates that the difference between the OLS and 2SLS estimates of  $\beta$  is not statistically significant. The results from the first stage of the 2SLS estimate indicate that the Protestant share variable is a valid instrument for human capital from schooling:

$$\log(H/L) = 0.78 \log(K/L) + 0.10 \log(\text{Prot Sh}) - 0.14 \text{ SSA} + 2.09 \quad R^2 = .85$$

$$(.07) \quad (.03) \quad (.23) \quad (0.79) \quad (\mathbf{9})$$

It is possible that both the OLS and the 2SLS estimates of  $\beta$  are biased upward if the OLS estimate exhibits simultaneity bias and the 2SLS estimate is biased upward by the higher income in the countries with a high Protestant share of the population. Columns (5) and (6) present estimates of the model when the countries

	TABLE 2.	Effect o	f human	capital	from	schooling	on national	income
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	(1)	(2)	(3)	(4)	(5)	(6)
Technique	OLS	OLS	2SLS	OLS	OLS	2SLS
Countries	36	36	36	36	36	36
Log(K/L)	0.48*	$0.47^{*}$	$0.50^{*}$			
	(0.05)	(0.05)	(0.07)			
Log(H/L)	0.23*	0.24*	$0.20^{*}$			
	(0.04)	(0.05)	(0.06)			
Log(K/Y)				$0.89^{*}$	0.92*	0.79
				(0.28)	(0.26)	(0.33)
Log(H/Y)				0.59*	0.42	0.80*
				(0.13)	(0.17)	(0.22)
Log(Protestant		-0.01			0.06	
Share)		(0.01)			(0.05)	
$R^2$	.94	.94	.94	.51	.53	.49
Implied $\alpha$	0.48	0.47	0.50	0.36	0.39	0.31
Implied $\beta$	0.23	0.24	0.20	0.24	0.18	0.31
Implied $\alpha$ + implied $\beta$	0.71	0.71	0.70	0.60	0.57	0.62

Note: Dependent variable is Log(national income/adult). White-adjusted standard errors in parentheses.

with a Protestant share of the population greater than 60% are excluded from the data set. This restriction eliminates 4 countries, leaving a sample of 57 countries and reducing the correlation between  $\log(H/L)$  and  $\log(\text{Prot Share})$  from 0.42 to 0.30 in the remaining sample. The OLS estimate of  $\beta$  in this subsample is identical to the estimate in the original sample. The 2SLS estimate of  $\beta$  in this subsample is higher than that in the original sample and statistically significant at a 3% level of confidence. There is no evidence that the four countries with very high levels of Protestant affiliation are driving the 2SLS estimate of the value of  $\beta$  in the larger sample of countries.

It is possible that the estimates of  $\alpha$  and  $\beta$  are biased if relevant explanatory variables are omitted from the income model. The very high  $R^2$  provides some confidence that no important variables are missing, but conceivably such omitted factors could be highly correlated with the two capital factors. This possibility is examined in Appendix C. The results provide evidence that no important factors are missing and that the estimates of  $\alpha$  and  $\beta$  are robust to the inclusion of other variables in the model.

Even though the estimates of  $\alpha$  and  $\beta$  are likely to be more accurate with a larger cross-country data set, the use of 61 countries to create the macro estimates rather than the 36 countries used for the micro estimates raises the possibility that the larger sample is not representative of the countries to be compared. Table 2 presents estimates of the MRW model using only the 36 countries to be compared. Columns (1)–(3) show the results for the MRW model in its basic form. In these

<sup>\*</sup>Significant at one percent level.

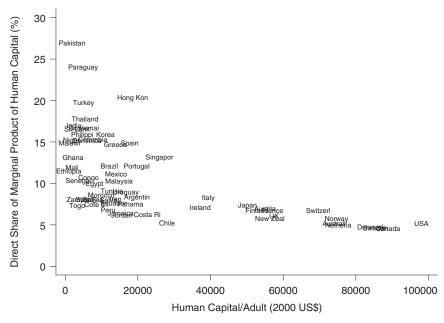


FIGURE 5. Direct share of macro marginal product of human capital—1990.

results the estimates of  $\alpha$  are unacceptably high (0.48–0.50), and the estimates of  $\beta$  (0.20–0.23) are lower than in the larger set of countries. One possible explanation for these results is the high correlation ( $\rho = 0.91$ ) between K/L and H/L in these data.

Columns (4)–(6) present the estimates for a common reduced form of the MRW model. These estimates of the effect of physical capital and human capital are more similar to the results with the larger data set. The estimates of the effect of physical capital are lower because of bias, as is expected when the model is estimated with K/Y instead of K/L [Cohen and Soto (2007)]. In the 2SLS estimates implied  $\alpha=0.31$  and implied  $\beta=0.31$ , the sum of whose values (0.62) is about 11% lower than their sum (0.70) in the direct form of the model. This test provides confirmation that the estimates of  $\alpha$  and  $\beta$  in the 61-country data set are representative of the relationship in the 36-country data set.

The direct share of the marginal product of human capital (Direct =  $\beta^2 Y/H$ ) calculated using the estimates for the 61-country data set is shown in Figure 5 for the 61 countries in the macro data set. Even though the statistical tests indicate that the OLS estimate of  $\beta = 0.31$  is unbiased, the 2SLS estimates of  $\alpha = 0.34$  and  $\beta = 0.36$  are preferred because of the existence of the independent evidence that  $\alpha = 0.35$ . For this reason the 2SLS estimates of  $\alpha$  and  $\beta$  are used to calculate the marginal product of human capital in the MRW model. This estimate of  $\beta$  is higher than estimates in earlier studies, which is likely a result of the larger income differences due to the inclusion of foregone student earnings, the use of data for

		Rate	of return	(%)	Share	e of invest	tment	Rate (%)
Country	Year	Primary	Second	Higher	Primary	Second	Tertiary	Average
Argentina	1989	8.4	7.1	7.6	0.50	0.33	0.17	7.8
Bolivia	1990	13	6.0	13.0	0.36	0.41	0.23	10.1
Brazil	1989	35.6	5.1	21.4	0.49	0.32	0.19	23.1
Chile	1993	8.1	11.1	14.0	0.41	0.44	0.15	10.3
Colombia	1989	20	11.4	14.0	0.52	0.27	0.21	16.4
Costa Rica	1989	11.2	14.4	9.0	0.44	0.20	0.36	11.0
Ecuador	1989	14.7	12.7	9.9	0.38	0.33	0.30	12.6
El Salvador	1990	16.4	13.1	8.0	0.47	0.27	0.26	13.3
Jamaica	1989	17.7	7.9	7.9	0.54	0.37	0.08	13.2
Mexico	1992	11.8	14.6	11.1	0.52	0.24	0.24	12.3
New Zealand	1991	$12.4^{a}$	12.4	9.5	0.45	0.41	0.15	12.0
Paraguay	1990	20.3	12.7	10.8	0.56	0.26	0.18	16.6
Philippines	1988	13.3	8.9	10.5	0.42	0.36	0.22	11.1
Spain	1991	7.4	8.5	13.5	0.62	0.29	0.09	8.3
ÚK	1986	8.6	7.5	6.5	0.40	0.49	0.10	7.8
Uruguay	1989	21.6	8.1	10.3	0.49	0.29	0.21	15.2
USA	1987	$10.0^{a}$	10.0	12.0	0.37	0.43	0.20	10.4

TABLE 3. Rates of return on investment from earnings studies: Full return method

human capital that has less measurement error than the data used in other studies, and the use of the human capital stock rather than the H/Y ratio in the model.

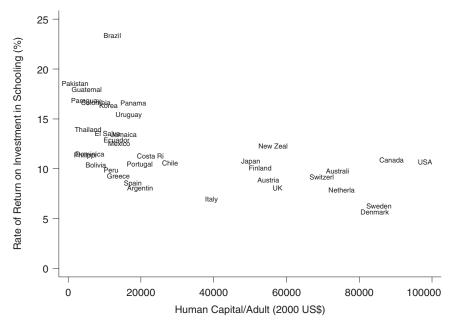
The rates overall exhibit a clear pattern of diminishing returns, which is particularly pronounced for countries with very little human capital. The variation in these rates is higher across countries with net human capital below \$30,000/adult (2000US\$), which could be because of greater measurement error in the data for these countries.

### 5. RATES OF RETURN IN WORKERS' EARNINGS STUDIES

Psacharopoulos and Patrinos (2004) summarize the rates of return on investment in schooling from national workers' earnings studies. They include "full return" private rates and social rates for all levels of schooling during the period 1986–1994 for 17 of the 61 countries included in the macro data set. I use the social rates of return in this analysis because they include public expenditures on schooling. These rates exhibit considerable variation by level of schooling within countries and at the same level of schooling across countries. These rates and the weighted average rate, which are actually the marginal products of schooling, are shown in Table 3.

Psacharopoulos and Patrinos (2004) summarize the Mincerian rates for 35 of the 61 countries included in the macro data set. In this data set the rates for Japan and Italy are outliers, in that Japan is unusually high (13.2%) and Italy is unusually

<sup>&</sup>lt;sup>a</sup>Unavailable rate for the primary level assumed to be the same as the rate at the secondary level.



**FIGURE 6.** Return on investment in schooling from earnings studies—1990.

low (2.7%). I investigated the sources for these rates and found that the rates are incorrect. Cohn and Addison's (1998) rates for Japan range from 4.4% to 13.2%, with a midpoint of 8.8%. Brunello and Miniaci's (1999) rate for Italy is 5.6%. I used these rates for these countries.

Of the 36 countries that had either a full return rate or a Mincerian rate, 16 countries had both rates. Two countries, Brazil and Jamaica, have pairs of rates that are outliers. Excluding these two countries, the remaining 14 countries have an average full rate of return (i.e., a marginal product) that is 1.19 times the average Mincerian rate. I use this factor to estimate the marginal product of schooling for the 19 countries that had only a Mincerian rate.

The resulting set of micro rates (actually marginal products of schooling) is shown in Figure 6 and is provided in Appendix B. Although there is a clear pattern of diminishing returns, there is also considerable variation in these rates. This variation is likely due in part to measurement error in the data on workers' earnings and school expenditures, and especially in the estimates of students' foregone earnings, which, as shown in Appendix A, can vary considerably between micro studies because of differences in estimation methodologies.

### 6. COMPARISON OF MACRO AND MICRO MARGINAL PRODUCTS

Table 4 provides a summary of the estimates of the micro and macro marginal product data for the 36 countries that had both rates. The average direct share of

Variable	Obs.	Mean	Std. dev.	Minimum	Maximum
Micro—Earnings studies	36	11.6	3.9	5.4	23.1
Macro—Direct—Low experience	36	9.8	5.6	4.3	26.6
Macro—Direct (adj)	36	10.7	6.1	4.7	29.2
Macro—Direct (adj) + External	36	19.6	11.2	8.6	53.4

TABLE 4. Summary characteristics for micro and direct macro rate-of-return data

the marginal product of human capital in the MRW model was 9.8%, whereas the average micro marginal product of schooling was 11.6%. The correlation coefficient between these two sets of rates is .57. A Student's t test indicates that the two data samples are correlated, with a level of significance above 1%.

The two sets of rates are not entirely comparable, because the workers' level of experience is not the same in the two data sets. Because of the growth in the world population over the period 1930–1970, the 1990 worldwide work force in the macro analysis had about 2.5 years less experience than it would have had in the absence of population growth.

The earnings patterns in Figure 2, combined with the 1990 distribution of schooling by level in Cohen and Soto (2007), indicates that a 2.5-year reduction in average experience (12.5% of 20 years) is likely to have reduced the aggregate earnings of the workers in the macro study by about 6% relative to those of workers with an average level of experience. Adjusting national income upward in each country to offset this decline in the average experience of the work force raises the mean of the direct macro rates to 10.7%. The associated adjusted macro rates that include the external effect of schooling on the labor wage (w) are also shown in the table. These rates are a 1.83 multiple  $[(1-\alpha)/\beta]$  of the adjusted direct macro rates. The mean of these rates is 19.6%.

The mean of the micro rates is 12% greater than the mean of the direct adjusted macro rates and 41% less than the mean of the direct + external adjusted macro rates. So the mean of the micro rates is consistent with the means of the adjusted macro rates, but the consistency of the means does not ensure that the two sets of rates are consistent across countries with different levels of human capital.

Figure 7 shows the individual estimates of the micro rates for the 36 countries and the fitted trend for these estimates, ordered by the national level of (net) human capital from schooling. The figure also shows the fitted trend for the direct share of the adjusted macro rates (i.e., the share that accrues directly to the recipient of schooling) and the fitted trend for the direct share plus the external share of the adjusted macro rates that accrues to labor for the same 36 countries. Given the considerable variation in individual micro and macro rates, the trends in the data are more meaningful than the individual rates.

Both the micro and the macro rates exhibit diminishing returns, but the relationship between these rates is quite different over the range of countries. In the countries with low levels of human capital from schooling, the trend for the micro

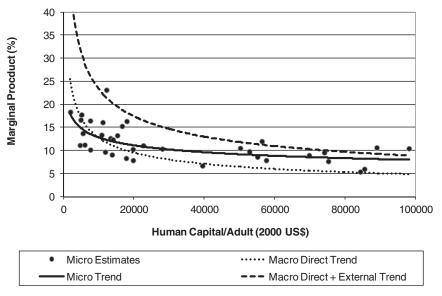


FIGURE 7. Macro and micro marginal products of schooling.

rates tracks the trend for the direct share of the macro rates. As the level of human capital from schooling rises, the trend for the micro rates rises above the trend for the direct share of the macro rates. For the countries with the highest level of human capital, the micro trend approaches the trend for the macro rates that include the sum of the direct share and the external share that accrues to labor.

The implication of the pattern of rates in Figure 7 is that in countries with low levels of human capital, the external marginal product of schooling in 1990 accrued equally to the entire work force, so it did not affect the micro estimates of the marginal product of schooling. In countries with human capital from schooling of \$100,000/adult, the external marginal product of schooling accrued to the more schooled workers, so the effect is included in the micro estimates of the marginal product of schooling.

How robust are these trends in the rates? The trend for the micro rates is quite robust, with a trend that is unlikely to exhibit much measurement error. Psacharopoulos and Patrinos (2004) present many estimates of rates of return on investment in schooling for more recent years in many countries, and they are consistent with these rates.

The trends in the macro rates are based on the upper bound MPH =  $(1 - \alpha)\beta Y/H$  and the lower bound MPH =  $\beta^2 Y/H$ . The value of  $\alpha$  is approximately 0.35 in all valid estimates of the MRW model, so it is stable. Because there are only two variables in the model, K/L and H/L, which are similar in magnitude and highly correlated with Y/L, the sum  $\alpha + \beta$  is extremely stable. As shown in Table 2, when the model is estimated as a function of K/Y, this (biased) sum tends to

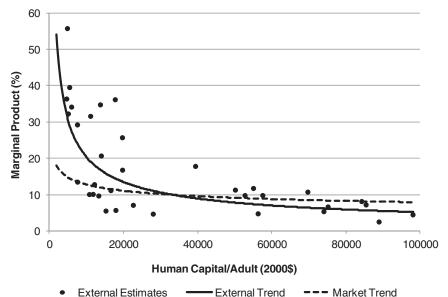


FIGURE 8. Market and external marginal products of schooling.

be 0.6, as it is in MRW's original estimates. The sum of the unbiased estimates is expected to be higher, closer to 0.7, as shown in Table 1. As a result, if  $\alpha = 0.35$ ,  $\beta$  must be approximately 0.35 as well.

In the current estimate of  $\beta$  in the MRW model, the econometric results are unaffected by the ratio of foregone earnings to direct expenditures because this ratio is constant across countries and the MRW model is estimated in log form. But this ratio does affect the estimate of H in the macro rates, so this is where the macro rates may encounter uncertainty. In the reconciliation this ratio is assumed to be 1.9 for all countries, but the data shown in Appendix I indicate that this ratio conceivably could be as low as 1.7 and as high as 2.1. This variation would change the macro rates by -1% at the upper bound and -0.5% at the lower bound. Variations of this order of magnitude in the macro marginal products would not invalidate the results of the reconciliation exercise. The micro and macro rates would still be consistent.

Even in countries with human capital/adult of \$100,000/adult, not all of the marginal external effect of schooling in 1990 was captured by schooled workers, because the  $\alpha$  share of the marginal product in the macro rates of return does not accrue to the work force. As a result, the nation's marginal product of human capital substantially exceeds the marginal product of schooling in workers' earnings studies in all countries. This external marginal product, the difference between the macro and micro marginal products, is shown in Figure 8. The external marginal product in 1990 was about 2/3 of the market rate (actually the marginal product of

schooling) in countries with high levels of human capital, but it was substantially above the market rate in countries with low levels of human capital.

The empirical results indicate that the decline in the marginal product of human capital due to diminishing returns is relatively slow. With the increase in the net human capital stock from \$50,000/adult to \$100,000/adult (2000 U.S.\$), the market rate in 1990 declined from about 10% to about 8.5%, whereas the total marginal product, including the external effects, declined from about 18% to about 14%.

The external marginal product includes the marginal product that accrues to physical capital and to labor. The external marginal product to labor alone, the difference between the share of the marginal product accruing to workers and the micro marginal product, is not shown in Figure 8. As can be inferred from the relationship in Figure 7, this rate was high in countries with little human capital in 1990 but diminished to a very low rate in countries with (net) human capital from schooling greater than \$90,000/adult (2000 U.S.\$).

Empirical studies of the external effect of human capital on physical capital and labor within countries are consistent with the results in Figures 7 and 8. Chi (2008) and Lopez-Bazo and Moreno (2008) present evidence that historically an increase in human capital raised investment in physical capital in China and Spain. Acemoglu and Angrist (2001) and Moretti (2004) present evidence that human capital has a small external effect on workers' salaries in the United States. Liu (2007) and Fleisher et al. (2010) present evidence that human capital has a large external effect on workers' salaries in China.

### 7. CONCLUSIONS

This paper presents the results of a test to determine whether Mankiw, Romer, and Weil's model of economic growth has a marginal product of human capital that is consistent with the estimates of the return on investment in schooling in micro studies. In the MRW model human capital has two external effects, one on physical capital and one on labor. Because workers receive income from human capital through the direct effect ( $\beta$ ) and through the external or spill-over effect ( $1-\alpha-\beta$ ) on labor, their total income is bounded by the direct effect on the low side and the sum of these two effects on the high side. If the MRW model is valid, independent estimates of the increase in workers' incomes that accompanies increased schooling should fall within these bounds.

After estimating the MRW model with a new set of human capital and national income data, I show that across 36 countries the fitted trends for the macro estimates of the marginal product of human capital in 1990 are consistent with the fitted trends for the micro estimates. In this comparison, the relationship between the micro and macro rates follows a well-defined pattern. In the lowest-income countries, the micro rate is equal to the direct share of the macro marginal product (i.e., the  $\beta$  share). The share accruing to workers increases as countries increase their level of human capital, until in the highest-income countries, the micro rate is almost equal to the direct share plus the external share of the marginal product

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that accrues to labor (i.e., the  $1-\alpha$  share). These results indicate that in 1990, as the stock of human capital increased, a higher share of the external returns from schooling accrued to the more educated component of the work force. The results indicate that physical capital and skill are complementary at all levels of human capital. They indicate that human capital and unschooled labor are also complementary, but the marginal effect of human capital on the marginal product of labor is small at high levels of human capital.

The empirical results in this analysis provide evidence that MRW's growth model is a valid representation of the growth process. The results also provide evidence that investment in human capital has large external effects on national income and that these effects are larger in countries with lower levels of human capital.

### **NOTES**

- 1. For example, hours worked per year vary across countries but are not included as a control variable.
- 2. Cohen and Soto (2007) obtain similar estimates of the model within countries between 1960 and 1990. Breton (2010) obtains similar estimates across countries for the period 1990–2000.
- 3. Mauro (1995) presented evidence that a measure of a country's bureaucratic efficiency, which has since been interpreted to measure either a country's institutions or its policies, affects its level of physical capital.
- 4. Although an increase in human capital from schooling raises the return on physical capital, in a market economy this increase is theoretical. As incremental schooling raises the marginal product of physical capital, investment in physical capital increases, which causes the marginal product to remain at the market rate of return. The share of national income accruing to physical capital continues to equal  $\alpha$ , but the higher income is then due to the increase in the stock of physical capital rather than to the increase in its marginal product.
- 5. The remaining share of the marginal product of human capital ( $\alpha$  MPH<sub>i</sub> =  $\alpha\beta Y_i/H_i$ ) accrues to physical capital.
- 6. Critics have challenged the methodology used to estimate Mincerian rates because it incorrectly assumes that the effects of schooling and experience on income are independent. Despite this conceptual deficiency, the method used to estimate the full return rates from the Mincerian rates in this analysis is still valid.
- 7. As described later, a dummy variable for the sub-Saharan African countries is included in the income model to control for the negative effects of worker morbidity and mortality on the effective human capital stock in these countries.
- 8. The foregone earnings ratio does not affect the estimate of the effect of human capital on national income  $(\beta)$  because the earnings ratio is constant across countries and the income model is estimated in log form.
- 9. The earnings data were obtained from Gomez-Castellanos and Psacharopoulos (1990), Hossain and Psacharopoulos (1994), Psacharopoulos and Velez (1994), and Psacharopoulos et al. (1994).
- 10. Financial depreciation rates for physical capital have a geometric pattern because the income generated by physical capital does not rise over time.
- 11. Hong Kong's national income is somewhat of an outlier relative to its level of human capital, no doubt because of its unique capability in 1990 to benefit financially from its relationship with China.
- 12. The data in PWT 6.2 are benchmarked to 1996 data, which provide a good basis for an analysis in 1990.

- 13. There has long been a hypothesis that the rise of capitalism is related to the Protestants' promotion of literacy and education to facilitate the study of the Bible [Means (1966) and Craig (1981)].
- 14. In a recent test of Weber's hypothesis, Becker and Wössmann (2009) show that in 453 counties in Prussia in 1871, incomes were higher in Protestant than in Catholic counties, but that after controlling for the effect of literacy, there is no significant difference in economic success between Protestant and Catholic counties.
- 15. Bernanke and Gurkaynak (2002) did not include students' foregone earnings in the national income data they used to estimate the share of national income accruing to physical capital. If they had included these foregone earnings, their estimate of the share of national income accruing to physical capital would have been slightly smaller.
- 16. Ciccone and Jarocinski (2010) do not address the endogeneity problem, but they replicate the analysis of Sala-i-Martin et al. (2004), who state that they chose their variables from early in the growth period partly to deal with the endogeneity problem.

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# APPENDIX A: RATIO OF FOREGONE EARNINGS TO NATIONAL EXPENDITURES ON SCHOOLING

I estimated the ratio of foregone earnings to cumulative national expenditures on schooling in 1990 using data from studies in nine countries. The data came from several sources. I obtained raw data for nine countries on direct costs and foregone earnings by level of schooling in the late 1960s from Psacharopoulos (1973). I weighted these data using the 1970 distribution of schooling by level provided in Cohen and Soto (2007) and an assumed schooling duration of six years for primary and secondary schooling and four years for university. The resulting estimates of the ratio of foregone earnings to direct expenditures are shown in Table A.1. These estimates range from 0.23 in Malaysia to 1.54 in Great Britain. The average ratio is 1.02. The enormous variation in these estimates is not a function of the level of national income.

Psacharopoulos's notes reveal that in some countries the foregone earnings data are not representative of all workers. In the United States the data are for white males only. In Great Britain foregone earnings at the university level are for science and engineering students. In Colombia workers were assumed to work full time for 50 weeks, with no period of unemployment. These differences are likely to lead to an overestimate of foregone earnings under actual working conditions in these three countries.

Tilak (1988) provides estimates of the ratio of foregone earnings to expenditures of schooling in India in 1979–1980. The upper bound is 0.72. He also provides Kothari's estimates of the ratio of foregone earnings to direct expenditures in India for 1960, which averaged 0.99. Both of these estimates are below the ratio of 1.12 for India calculated from

	•		
	Estimated <sup>a</sup>	Tilak (1988)	Kendrick (1976)
Chile	0.69		
Colombia	0.79		
Great Britain	1.54		
India	1.12	0.86	
Malaysia	0.23		
Mexico	1.53		
New Zealand	1.03		
South Korea	1.27		
United States	0.99		0.90
Average	1.02		

TABLE A.1. Ratios of foregone earnings to direct schooling costs

 $<sup>^</sup>a$ Estimated from data on direct costs and foregone earnings in Psacharopoulos (1973).

the data in Psacharopoulos (1973). The average ratio for the two comprehensive studies in India is 0.86.

Kendrick (1976) estimated the total capital investment in the United States in 1969. He estimated that the ratio of foregone earnings to direct expenditures was 0.90. This estimate also is below the ratio of 0.99 for the United States calculated from the data in Psacharopoulos (1973).

Because foregone earnings are overestimated in Psacharopoulos's data, the average ratio of foregone earnings to direct expenditures in the nine countries is below 1.02. The average ratio in the two comprehensive studies in India and the United States is 0.88. Based on this review of the data in these studies, I estimate that that foregone earnings were approximately 90% of direct expenditures in all countries.

# APPENDIX B: DATA USED IN THE ANALYSIS

The key data used in the analysis and some of the data used to create these data are provided in Table A.2. The national income and human capital data are in 2000 U.S.\$.

**TABLE A.2.** Data used in the analysis

Country	<i>Y/L</i> \$/adult	H/L \$/adult	School fin mult ratio	Model MPH %	Labor growth %	Adjusted MPH %	Mincer return %	Micro MPS %
Argentina	12,305	19,766	1.39	22.4	1.6	24.6	10.3	7.8
Australia	27,834	74,055	1.58	13.5	2.0	14.8	8.0	9.5
Austria	28,286	55,069	1.49	18.5	0.4	20.3	7.2	8.6
Bolivia	4,561	7,637	1.34	21.5	2.2	23.6	10.7	10.1
Brazil	11,096	12,214	1.33	32.7	2.9	35.8	14.7	23.1
Canada	29,461	88,933	1.55	11.9	2.1	13.1	8.9	10.6
Chile	10,622	28,037	1.40	13.6	2.2	14.9	12.0	10.3
Colombia	8,793	7,610	1.34	41.6	2.9	45.6	14.0	16.4
Congo-Rep	5,163	6,703	1.32	28.9				
Costa Rica	10,400	22,667	1.33	16.5	3.3	18.1	8.5	11.0
Cote d'Ivoire	4,739	9,014	1.32	19.9	3.9			
Denmark	28,730	84,260	1.53	12.3	0.8	13.5	4.5	5.4
Dominican Rep	6,921	6,034	1.33	41.3	3.0	45.3	9.4	11.2
Ecuador	7,528	13,352	1.35	20.3	2.9	22.2	11.8	12.6
Egypt	6,033	8,081	1.31	26.9				
El Salvador	6,418	10,845	1.32	21.3	2.5	23.4	7.6	13.3
Ethiopia	896	1,036	1.32	31.1				
Finland	26,090	52,791	1.44	17.8	0.9	19.5	8.2	9.8
France	28,326	57,044	1.44	17.9				
Ghana	2,143	2,159	1.32	35.7				
Greece	15,288	13,792	1.39	39.9	1.0	43.7	7.6	9.0
Guatemala	6,522	5,154	1.31	45.6	3.0	49.9	14.9	17.7
Hong Kong	28,855	18,620	1.37	55.8				

TABLE A.2. (Continued.)

Country	<i>Y/L</i> \$/adult	H/L \$/adult	School fin mult ratio	Model MPH %	Labor growth %	Adjusted MPH %	Mincer return %	Micro MPS %
India	3,112	2,413	1.32	46.4				
Iran	8,375	13,910	1.32	21.7				
Ireland	19,460	37,106	1.44	18.9				
Italy	24,425	39,449	1.39	22.3	0.8	24.4	5.6	6.7
Jamaica	7,250	15,293	1.37	17.1	1.5	18.7	28.8	13.2
Japan	27,617	50,190	1.55	19.8	1.6	21.7	8.8	10.5
Jordan	7,088	15,669	1.35	16.3				
Korea, Rep	13,451	11,150	1.40	43.4	2.7	47.6	13.5	16.1
Malawi	1,323	1,258	1.32	40.6				
Malaysia	11,431	14,827	1.34	27.8				
Mali	1,703	2,050	1.32	32.3				
Mexico	11,731	14,052	1.35	30.1	2.9	32.9	7.6	12.3
Morocco	6,366	9,977	1.32	23.0				
Netherlands	27,099	75,178	1.50	13.0	1.4	14.2	6.4	7.6
New Zealand	23,822	56,321	1.52	15.2	1.7	16.7	10.3	12.0
Niger	2,019	1,868	1.32	41.4				
Norway	31,529	74,667	1.54	15.2				
Pakistan	3,998	1,944	1.31	74.0	2.4	81.1	15.4	18.3
Panama	9,998	17,969	1.35	20.0	2.7	22.0	13.7	16.3
Paraguay	8,981	4,904	1.34	65.9	2.5	72.3	11.5	16.6
Peru	5,919	11,853	1.35	18.0	2.8	19.7	8.1	9.6
Philippines	5,652	4,705	1.35	43.3	2.9	47.4		11.1
Portugal	17,942	19,729	1.33	32.7	0.7	35.9	8.6	10.2
Senegal	2,823	3,911	1.32	28.0				
Singapore	25,801	25,985	1.37	35.7				
Spain	20,032	17,803	1.39	40.5	1.1	44.4	7.2	8.3
Sri Lanka	4,152	3,306	1.34	45.2				
Sweden	28,411	85,441	1.53	12.0	0.7	13.1	5.0	6.0
Switzerland	34,665	69,732	1.63	17.9	1.2	19.6	7.5	8.9
Syria	3,039	5,067	1.32	21.6				
Thailand	7,413	5,504	1.32	48.5	3.0	53.1	11.5	13.7
Togo	1,804	3,333	1.32	19.5				
Tunisia	8,756	12,898	1.31	24.4				
Turkey	7,604	5,063	1.32	54.1				
UK	25,680	57,567	1.54	16.1	0.4	17.6	6.8	7.8
Uruguay	11,112	16,636	1.37	24.0	0.9	26.4	9.7	15.2
USA	36,912	98,073	1.60	13.5	1.4	14.9	10.0	10.4
Zambia	2,132	3,802	1.33	21.5				

## APPENDIX C: OMITTED VARIABLES ANALYSIS

Hundreds of cross-country studies have tried to determine which factors affect economic growth. Levine and Renelt (1992) examined over 50 variables that had been shown to be correlated with economic growth and showed that few are robust to changes in the accompanying set of variables in the growth model. Since then, numerous researchers have used Bayesian techniques in an effort to determine whether some variables are more robust than others. Depending on how they structure their priors and which data sets and time periods they analyze, they come to different conclusions. Ciccone and Jarocinski (2010) conclude that variables purported to be robustly correlated with growth are not robust to changes in the cross-country data sets.

One serious limitation of Ciccone and Jarocinski's study and the others that they replicate is that they do not control for endogeneity bias. They attempt to limit this bias by only selecting explanatory variables from early in the growth period, but in the process they introduce specification error or measurement error into their analysis. <sup>16</sup> For example, in the dynamic version of the MRW model, the rates of investment in physical capital and human capital are the constant (or at least the average) rates over the growth period. When the data for these variables are taken from early in the period to minimize endogeneity bias, these variables are at least poorly measured. As a result, when the neoclassical variables are found not to be robust, it is not clear if this is because they are invalid or because they are mismeasured.

Mirestean and Tsangarides (2009) employ a Bayesian technique to examine whether a large number of variables are robust, and they attempt to control for endogeneity by using lagged variables and differences in the lagged variables as instruments. As will be discussed later, it is not clear that they succeeded in eliminating endogeneity bias in all of their estimated coefficients, but they at least made an effort to do so, and they managed to identify some variables that according to their methodology are robust. In this Appendix I examine whether the estimates of  $\alpha$  and  $\beta$  in the MRW model are robust to the addition of these variables to the model.

Mirestean and Tsangarides performed several different analyses. Overall they found the strongest evidence for the robustness of initial income, investment, population growth, and life expectancy. They found less strong evidence for debt, trade openness, and inflation.

They found only weak evidence for the robustness of education, but the variable they created for education was the average stock of primary and secondary schooling. It is not clear why they excluded tertiary education from their variable, but its exclusion undermines the validity of their results, because it eliminates most of the variation in education across high-income countries. One common problem with the Bayesian analyses is that although they examine a vast array of variables, invariably some of these variables are not ideally specified.

Three of the four most robust variables in their analysis (initial income, investment, and population growth) are the standard neoclassical variables, which implicitly are already included in the MRW model. This leaves debt, trade openness, inflation, and life expectancy as the four additional variables to be tested. Inflation and debt are expected to have a negative effect, whereas life expectancy and trade openness are expected to have a positive effect. The specific variables tested are the net government debt–to–GDP ratio in 1990, the trade openness variable in 1990, log(1+ the GDP deflator) for 1990, and life expectancy at birth

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Technique	OLS	OLS	OLS	2SLS	2SLS	2SLS	OLS	2SLS
Countries	61	61	61	61	61	61	57	32
Log(K/L)	0.38*	0.39*	0.41*	0.34	0.33*	0.33	$0.47^{*}$	0.40
	(0.06)	(0.06)	(0.07)	(0.14)	(0.12)	(0.14)	(0.15)	(0.16)
Log(H/L)	0.31*	0.22*	0.29*	0.36	0.31	0.37	0.22	0.32
	(0.06)	(0.05)	(0.06)	(0.15)	(0.15)	(0.15)	(0.09)	(0.13)
Sub-Saharan	-0.28*	-0.02	-0.28*	-0.29*	-0.10	-0.29*	-0.05	-0.08
Africa	(0.11)	(0.15)	(0.11)	(0.11)	(0.17)	(0.10)	(0.19)	(0.19)
Debt to GDP							-0.42	-0.52
ratio							(0.39)	(0.45)
Trade		-0.09	-0.06		-0.07	-0.04	-0.12	-0.10
openness		(0.06)	(0.07)		(0.08)	(0.09)	(0.07)	(0.07)
Inflation		-0.06	-0.08*		-0.06	-0.07	-0.00	0.23
		(0.03)	(0.03)		(0.03)	(0.04)	(0.56)	(0.58)
Life		0.02			0.01		0.01	0.01
expectancy		(0.01)			(0.01)		(0.02)	(0.02)
$R^2$	.95	.96	.96	.95	.96	.95	.98	.98

TABLE A.3. Effect of human capital from schooling on national income

Note: Dependent variable is Log(national income/adult). White-adjusted standard errors in parentheses. \*Significant at 1% level.

in 1990. The data for these variables were obtained from IMF statistics (debt), PWT 6.2 (openk), and the World Bank's World Development Indicators.

The empirical results for the model are shown in Table A.3. Column (1) shows the OLS estimate of the MRW model. Column (2) shows the results for this model, including all of the additional variables except debt. Debt was excluded initially because data were available in 1990 for only 32 of the 61 countries. When the three variables are included in the model, the estimates of  $\alpha$  and  $\beta$  are robust, but  $\beta$  is smaller (0.22 vs. 0.31) than in the standard model. The coefficient on trade openness has the wrong sign and is not significant, but the coefficients on the other two variables have the correct signs and are statistically significant.

Column (3) shows the OLS results without life expectancy, which shows that the decline in  $\beta$  in column (2) is due almost entirely to the inclusion of this variable. Acemoglu and Johnson (2007) have examined the effect of life expectancy on economic development, and they show that although it reduces population growth, there is no evidence that it raises income per capita. It seems likely that life expectancy is endogenous in this income model, because life expectancy rises with national income. It is also highly correlated with human capital ( $\rho=0.83$ ), and conceptually it seems more likely that higher human capital raises life expectancy rather than the reverse.

Columns (4)–(6) show the 2SLS estimates for the same models. The estimates of  $\alpha$  and  $\beta$  continue to be robust, and in these results the magnitude of  $\beta$  is less affected by the inclusion of life expectancy in the model. The estimate of  $\beta$  declines from 0.36 to 0.31 when the other variables are added to the model.

Columns (7) and (8) show the results when the debt to GDP ratio is added to the model. The estimated coefficient on debt has the expected sign, but it is not statistically significant.

Although the sample includes only 32 countries, the estimates of  $\alpha$  and  $\beta$  continue to be robust. In the 2SLS estimates, the value of  $\beta$  declines to 0.32.

Overall, in the various models, the estimates of  $\alpha$  and  $\beta$  are consistently positive and statistically significant at the 5% level or higher. Only the inclusion of life expectancy reduces the estimate of  $\beta$  in a nontrivial way. Acemoglu and Johnson's (2007) evidence indicates that life expectancy is not a determinant of economic growth, so its ability to reduce the magnitude of  $\beta$  is likely to be a result of endogeneity bias. The inclusion of the other variables in the income model has virtually no effect on the estimates of  $\alpha$  and  $\beta$  in the regressions.

# APPENDIX D: A COMPARISON OF THE NET HUMAN CAPITAL STOCK TO AVERAGE SCHOOLING ATTAINMENT

This Appendix presents the estimates of the national income model in equation (8) using Cohen and Soto's (2007) and Barro and Lee's (2001) estimates of average schooling attainment. These estimates can be compared with the estimates in Table 1 to investigate whether the data on net human capital yield more acceptable estimates of the national income model than the schooling attainment data. Because of differences in the countries for which data are available, slightly fewer countries are used in these regressions than in the results shown in Table 1.

Figure A.1 shows the relationship between  $\log(H/L)$  and Cohen and Soto's (2007) estimates of schooling attainment in the population from 15 to 64 in 1990. This relationship is linear because investment per year of schooling rises at an increasing rate as average years of schooling rises [Breton (2010)]. Across the countries in the data set, average years of schooling vary by a factor of 17, whereas net human capital per adult varies by a factor of 100. For this reason, in the income model,  $\beta \log(H/L)$  is represented by  $\theta$  Attainment:

$$\log(Y/L)_i = c + \alpha \log(K/L)_i + \theta \operatorname{Attainment}_i + \gamma \operatorname{SSA}_i + \varepsilon_i.$$
 (D.1)

This log-linear relationship between income and schooling attainment is consistent with the standard Mincerian relationship.

Table A.4 presents the results from the various estimates of the model. The 2SLS estimates utilize the log of the Protestant share of the population as an instrument. The implied values of  $\beta$  shown in the table are estimated using a conversion factor to relate the effect of average schooling attainment to the effect of net human capital from schooling. For the Cohen and Soto (2007) 1990 data, a regression indicated that  $\log(H/L) = 0.32$  Attainment, whereas for the Barro and Lee (2001) 1990 data,  $\log(H/L) = 0.38$  Attainment.

The empirical results for the standard income model are typical of the statistical estimates in earlier studies, which provide inconsistent estimates of  $\alpha$  that are greater than physical capital's share of national income and estimates of  $\beta$  that are relatively small. The similarity of the OLS and the 2SLS estimates of the effect of average attainment indicates that these unsatisfactory estimates of  $\alpha$  are not due to simultaneity bias.

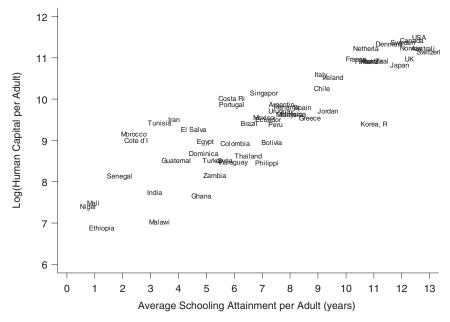


FIGURE A.1. Log(human capital per adult) vs. average schooling attainment in 1990.

TABLE A.4. Effect of human capital from schooling on national income

	Coh	Cohen and Soto (2007) data				Barro and Lee (2001) data				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)		
Technique	OLS	OLS	2SLS	2SLS	OLS	OLS	2SLS	2SLS		
Countries	56	56	56	56	58	58	58	58		
Log(K/L)	0.55*		0.52*		0.58*		0.54*			
	(0.07)		(0.07)		(0.06)		(0.08)			
Log(K/Y)		0.26		0.32		0.30		0.32		
		(0.16)		(0.15)		(0.15)		(0.14)		
Average	0.05	$0.17^{*}$	0.06	0.15*	0.05	$0.20^{*}$	0.07	0.20*		
Attainment	(0.02)	(0.02)	(0.03)	(0.03)	(0.03)	(0.02)	(0.04)	(0.03)		
Sub-Saharan	-0.18	-0.74*	-0.18*	-0.81*	-0.30	-0.78*	-0.29	-0.80*		
Africa	(0.19)	(0.25)	(0.20)	(0.28)	(0.14)	(0.18)	(0.15)	(0.21)		
$R^2$	.93	.81	.93	.81	.93	.78	.93	.78		
Implied $\alpha$	0.55	0.21	0.52	0.24	0.58	0.23	0.54	0.24		
Implied $\beta$	0.16	0.42	0.19	0.36	0.13	0.41	0.21	0.40		

Note: Dependent variable is Log(national income/adult). White-adjusted standard errors in parentheses.

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Table A.4 also presents the results for a reduced form of the income model, in which the model is converted to a function of the capital/output ratio (K/Y):

$$\log(Y/L)_i = c + (\alpha/1 - \alpha)\log(K/L)_i + (\theta/1 - \alpha) \operatorname{Attainment}_i + (\gamma/1 - \alpha)\operatorname{SSA}_i + \varepsilon_i.$$
(D.2)

The implied estimates of  $\alpha$  are much smaller in this model and are considerably smaller than the independent estimates of  $\alpha$  (0.35). In these models the estimates of  $\beta$  are much larger. The inconsistency in these estimates of  $\alpha$  and  $\beta$  across different versions of the income model apparently occurs because average schooling attainment is an inaccurate measure of human capital and is highly correlated with physical capital. The correlation between  $\log(H/L)$  and schooling attainment is .88 for the Cohen and Soto (2007) data and .84 for the Barro and Lee (2001) data in 1990. This is relatively high, but similar to the correlation between schooling attainment and  $\log(K/L)$ , which is .84 for the Cohen and Soto data and .81 for the Barro and Lee data.

It is clear from the pattern of the data in Figure A.1 that levels of human capital per adult (adjusted for purchasing power parity) vary considerably among countries with the same level of schooling attainment and that this variation is greater in countries with lower levels of average schooling attainment. Clearly some countries invest much more per year of schooling than other countries. If a nation's true human capital is a function of its cumulative investment in schooling, then average schooling attainment is an inaccurate measure of a nation's level of human capital in countries with relatively low levels of human capital.