The Role of Permanent Income and Demographics in Black/White Differences in Wealth*

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Abstract

We explore the extent to which the large race gap in wealth can be explained with properly constructed income and demographic variables. In some instances we explain the entire wealth gap with income and demographics provided that we estimate the wealth model on a sample of whites. However, we typically explain a much smaller fraction when we estimate the wealth model on a black sample. Using sibling fixed-effects models to control for intergenerational transfers and the effects of adverse history, we find that these factors are not likely to account for the lower explanatory power of the black wealth models. Our analysis of growth models of wealth suggests that differences in savings behavior and/or rates of return play an important role.

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I Introduction

The wealth gap between whites and blacks in the United States is much larger than the gap in earnings (see e.g. Menchik and Jianakoplos 1997). The gap in wealth has implications for the social position of African Americans that go far beyond its obvious implications for the consumption levels that households can sustain. This is because wealth is a source of political and social power, influences access to capital for new businesses, and provides insurance against fluctuations in labor market income. It affects the quality of housing, neighborhoods, and schools a family has access to as well as the ability to finance higher education. The fact that friendships and family ties tend to be within racial groups amplifies the effect of the wealth gap on the financial, social, and political resources available to blacks relative to whites.

The first question we ask in this paper is: To what extent can the large race gap in wealth be explained with income and demographic variables? Several previous studies (see Scholz and Levine (2002) for an up to date survey of the literature) have investigated the sources of the black/white wealth gap. We improve on these studies in a number of ways. First, the existing studies do not use an adequate measure of permanent income, which is a key determinant of wealth. Due to data limitations Smith (1995) and Avery and Rendall (1997) base their wealth models on current income alone rather than on current and permanent income. Blau and Graham (1990) and Menchik and Jianakoplos (1997) decompose income into current income and permanent income, where permanent income is the component that is predictable given race, sex, age, education, health status, number of children, and geographic location. Since wealth is a nonlinear function of income, however, use of the within-cell variation is necessary to precisely estimate wealth models. Moreover, because high-income individuals tend to have disproportionately large wealth holdings, failure to accurately measure the tails of the distribution of permanent income might lead to incorrect estimates of the contribution of permanent income to the wealth gap. We address this issue by using panel data from the Panel Study of Income Dynamics (PSID) to construct a measure of permanent income. Second, previous studies control for current
demographic variables such as marital status and presence of children but not for demographic histories which should influence current wealth through past savings. We address this by constructing marriage histories and child bearing and rearing histories. Third, to account for the fact that wealth is frequently zero or negative and has a highly skewed distribution and that demographic variables are likely to interact with income variables, we use both mean and median regression and work with models of the wealth/permanent income ratio as well as the level and log of wealth. A fourth problem arises from the substantial differences between the white and black distributions of income and demographics in combination with the fact that relatively flexible functional forms must be used to capture the nonlinearity of wealth in income (see also Barsky, Bound, Charles and Lupton 2002). These two features make it difficult to accurately estimate race-specific wealth models over the full range of income. We address this “nonoverlap” problem in several ways and check that our main results are robust to it.

With our improved income and demographic variables, we typically explain more of the wealth gap than previous studies (Section IV). In the case of single men and single women, for example, we can explain the entire race gap in the level of wealth with income and demographics provided that we estimate the wealth equation on the white sample. At the same time, we confirm earlier results that suggest large disparities in the explanatory power of the wealth models of whites and blacks. While income and demographics account for 79 percent of the wealth gap in the case of couples when we use the wealth mean regression model for whites, we explain only 25 percent when we use the wealth model for blacks. When we use the ratio of wealth to permanent income as our measure of wealth, we again explain a much larger fraction of the wealth gap using the coefficients for whites than the ones for blacks, although for couples and single females the fraction explained is smaller than in the case of the level of wealth. Our results indicate that race differences in self-employment patterns are a significant part of a full explanation of black/white wealth differences, although causality is ambiguous. We also show, in contrast to the analysis of Barsky et al. (2002), that these findings are not due to a lack of overlap between the white
The fact that we can explain most and in some cases all of the wealth gap with the white but not with the black wealth model leads us to focus on a second question: Why are the wealth holdings of blacks less sensitive to income and in demographics than the wealth holdings of whites? Starting with Blau and Graham (1990), some researchers have hypothesized that differences in inter vivos transfers and inheritances play a major part in explaining the wealth gap. In Section V, we propose a theoretical model of intergenerational linkages to show how the legacy of discrimination could lead to a link between wealth and income that is stronger for whites than for blacks. We study this possibility by relating differences among siblings in current and permanent income and demographics to differences in wealth. Using sibling comparisons largely neutralizes the effects of disparities between whites and blacks in inter vivos transfers and inheritances and provides a way of controlling for the effects of an adverse history on the relative position of blacks. Our study is the first (to our knowledge) sibling fixed-effects analysis of wealth holdings and is of independent interest. With fixed-effects coefficient estimates from the white sample we can explain 104 percent of the race gap in the wealth level, while the corresponding value based on the model for blacks is only 49 percent. Our sibling results thus confirm that wealth holdings are much less strongly related to income and demographic variables among blacks than among whites. They are also consistent with evidence in Altonji and Villanueva (2001) based on both the PSID and the survey of Asset and Health Dynamics (AHEAD). This suggests that little of the difference between whites and blacks in the effect of income on wealth is due to differences in inter vivos transfers and inheritances.

Other explanations for the racial difference in the sensitivity of wealth holding to income and demographic variables are that savings rates or the rates of return to assets differ. In Section VI, we look at the combined effect of these two factors by estimating models for the growth of wealth as a function of income and demographic factors. In the case of couples, income and demographic factors explain 74 percent of the race gap in the
growth of wealth when we use the growth model for whites but only 49 percent when we use the growth model for blacks. The discrepancy is even larger (84 percent versus 30 percent) when we study growth in the ratio of wealth to permanent income. These findings suggest that difference in savings behavior or rates of return on assets play a key role in explaining the wealth gap. However, the results are not entirely conclusive for reasons we discuss below.

II Econometric Models and Wealth Decompositions

Let $i = 1, \ldots, N^j$ index individuals or couples and $j$ demographic groups, where $j$ is $b$ for blacks and $w$ for whites. Let $W^j_i$ denote wealth, $Y^j_i$ a vector of income variables (including current income $y_{it}$ and permanent income $y_i$), and $X^j_i$ a vector of demographic variables. Since we have panel data on wealth, a number of variables in addition to $y_{it}$ depend on the calendar year $t$ as well as $i$, but we leave the time subscript implicit.

Our basic wealth model specifies the level of wealth to be linear in the income and demographic variables and is given by

\begin{align}
W^w_i = \alpha^w_0 + Y^w_i \alpha^w + X^w_i \beta^w + \epsilon^w_i,
\end{align}

\begin{align}
W^b_i = \alpha^b_0 + Y^b_i \alpha^b + X^b_i \beta^b + \epsilon^b_i,
\end{align}

where $\alpha^w_0$, $\alpha^w$, and $\beta^w$ are the regression intercept and slope parameters for whites, $\epsilon^w_i$ is the error term, and $\alpha^b_0$, $\alpha^b$, $\beta^b$, and $\epsilon^b_i$ are the corresponding parameters and error term for blacks. We define $\alpha^j_0 + Y^j_i \alpha^j + X^j_i \beta^j$ to alternatively be the conditional mean or the conditional median regression function. Separate sets of regressions are specified for couples, single males, and single females, so the slopes and intercepts depend on sex and marital status as well as on race.

Given that the level of wealth is used as the dependent variable, equations (1) and (2) impose additive separability between the income and demographic variables. We work with the levels of wealth because much of the public discussion is couched in terms of levels and
because of the large fraction of the population with zero or negative wealth. However, such level models imply that wealth is additively separable in income and demographic variables, which is unlikely. Consequently, we also work with the log of wealth and the ratio of wealth to permanent income of the head as the dependent variable. The log model implicitly allows for multiplicative interactions in the equation for the level of wealth. However, the log specification is poorly suited as a wealth model because wealth holdings are frequently close to zero or even negative. We deal with this problem by truncating these values prior to taking logs. The ratio specification avoids this problem altogether and implies that the effects of demographic variables are proportional to permanent income.

Our choice of control variables is driven largely by previous studies of wealth differences, although our specifications of permanent income and marriage and child bearing and rearing histories are more elaborate. We chose to start with and stick to a fairly broad set of variables rather than “fish around” for a shorter list of variables that are statistically significant.

In the case of couples, the income controls included in $Y_i^j$ are current family non-asset income, permanent income of the husband, permanent income of the wife, the squares of current income, head’s permanent income, and wife’s permanent income, as well as the products of current family income with the head’s and the wife’s permanent income. The income controls in the case of singles are similar except that spouse variables are excluded.

For single men and women our demographic controls are region (4 categories), residence in an standard metropolitan statistical area, number of kids in the family unit, a dummy for whether there is at least one kid in the family unit, the number of dependents outside the family unit and a dummy for whether there is at least one dependent outside the family unit, number of siblings, number of marriages, whether the recent marriage ended in widowhood, whether the most recent marriage ended in divorce, the number of children born or adopted, and dummies for health status only poor or fair, education (6 categories), self employed, and the year of the wealth observation. In the case of couples we include health, education, self employment, number of siblings, number of marriages, and
number of children born or adopted measures for both the head and the spouse. We also include spouse’s annual hours worked. In addition, we include fourth-order polynomials in the age of the husband and in the age of the wife (centered at 40) in \( X_{ij} \). Because we estimate wealth models by demographic group, our age controls implicitly account for group differences in life expectancy.

It is useful to discuss our choice of variables in light of economic theories of savings and wealth (see Browning and Lusardi (1996) for a survey). The basic life cycle model of consumer behavior augmented to include a bequest motive suggests that wealth at a point in time will be influenced by the level and timing of earnings over the lifecycle, rates of time preference, rates of return, and consumption needs of the household. This suggests a role for both current and permanent earnings, powers of age, and various demographic variables that may influence the level and timing of consumption needs and perhaps are related to rates of time preference. Some of the variables we include, such as health status, are likely to operate through several of these channels.\(^1\) We do not have direct measures of rates of return. Recent work emphasizes precautionary savings and risk aversion, suggesting a role for measures of risk exposure, tolerance, insurance and ability to borrow against future income. We do not have direct measures of these factors either. Some of the race differences in the effects of variables that we include may, however, be related to these factors.

We evaluate the explanatory power of our wealth models using a slightly modified regression decomposition (Blinder 1973, Oaxaca 1973) that allows for median regression models and accommodates the use of population weights in computing the wealth gap. Let \( \{\omega_{ij}\}_{i=1}^{N_j} \) denote a set of population weights such that \( \omega_{ij} > 0 \) and \( \sum_{i=1}^{N_j} \omega_{ij} = 1 \). (See Section III for details on how the weights are constructed.) Let \( Z_i^j = (1, Y_i^j, X_i^j) \) and \( \theta^j = (\alpha_0^j, (\alpha^j)', (\beta^j))' \). Equations (1) and (2) can be written as

\[
W_i^j = Z_i^j \theta^j + \epsilon_i^j, \quad j = w, b,
\]

where the definition of \( \theta^j \) and \( \epsilon_i^j \) depends on whether we use mean or median regression.
Let $\hat{W}_j = \sum_{i=1}^{N_j} \omega_j^i(Z_i^j \hat{\theta}^j)$ denote the mean of the predictions for individuals in demographic group $j$, where $\hat{\theta}^j$ is an estimate of $\theta^j$. For a given family type, say couples, write

$$\hat{W}_w - \hat{W}_b = \sum_{i=1}^{N_w} \omega_w^i(Z_i^w \hat{\theta}^w) - \sum_{i=1}^{N_b} \omega_b^i(Z_i^b \hat{\theta}^b)$$

$$= \left\{ \sum_{i=1}^{N_w} \omega_w^i(Z_i^w \hat{\theta}^b) - \sum_{i=1}^{N_b} \omega_b^i(Z_i^b \hat{\theta}^b) \right\} + \left\{ \sum_{i=1}^{N_w} \omega_w^i(Z_i^w \hat{\theta}^w) - \sum_{i=1}^{N_w} \omega_w^i(Z_i^w \hat{\theta}^b) \right\}.$$ 

The first term measures what the contribution to the wealth gap of race differences in the explanatory variables would be if the relationship between wealth and the explanatory variables was given by $\hat{\theta}^b$, the coefficient vector for blacks. The second term evaluates the contribution to the wealth gap of differences between whites and blacks in the wealth coefficients using the distribution of the explanatory variables for whites. The first term thus represents the part of the wealth gap $\hat{W}_w - \hat{W}_b$ that is “explained” by differences between blacks and whites in the explanatory variables, while the second term represents the unexplained part of the wealth gap. We contrast the decomposition based on the above equation with the decomposition the coefficient vector for whites to measure the part of the wealth gap that is explained by differences in income and demographics.

### III Data

The database is constructed from the main PSID files, the supplemental wealth files, the marriage history file, and the childbirth and adoption file. The PSID is based on a random sample of U.S. households in 1968 and a separate low income sample. The households were interviewed annually, providing many years of income data and long demographic histories for the panel members. We combine the main PSID files with the marriage history file and the childbirth and adoption file to create a demographic history for each individual that describes past and present marriages and child bearing and rearing. We complement the demographic histories with a rich set of current demographic variables.

Wealth data were collected in 1984, 1989, and 1994. The PSID contains a full set of
variables only for household heads ("heads") and their spouses ("wives"). For this reason our analysis is based on persons who were either a head or a wife in one of the three years for which wealth surveys were conducted. We pool the data from the 1984, 1989, and 1994 wealth surveys.

We use our samples without weighting when estimating wealth models. When computing descriptive statistics and wealth decompositions we use the family core weights that are supplied with each wave of the PSID to make our estimates nationally representative. We assign equal weight to each of the three wealth surveys.

Non-asset family income is our measure of current income. We exclude asset income because it may be affected by prior transfers and because it is a function of wealth. Our measure of wealth includes home equity, but it excludes Social Security and pension wealth. These are important exclusions, and it would be desirable to incorporate them into future research. The income and wealth measures are deflated to 1989 US$ using the CPIU.

Throughout, the log wealth (log current income) measures are constructed by taking the log of the respective values truncated at 50 (1,000). This allows us to keep observations with zero or negative wealth (current income) in our sample. Our results for levels, ratios and logs are not sensitive to excluding observations with wealth below 50.

A Measuring Permanent Income

We use the panel data on individuals to construct two measures of permanent income. They are based on the regression model

\[
y_{it} = X_{it} \gamma + \epsilon_{it},
\]

where \(y_{it}\) is either the level or the log of nominal non-asset family income of person \(i\) in year \(t\). (Recall that, given the level of current income \(y_{it}\), the log of current income is defined as \(\ln \max \{y_{it}, 1000\}\).) \(X_{it}\) consists of a fourth-order polynomial in age (centered at 40), a marital status dummy, an indicator for children, the number of children, and a set of
year dummies. Define $e_{it}$ to be the sum of an individual-specific effect and an idiosyncratic error term, $e_{it} = v_i + u_{it}$, and assume that the serial correlation in $u_{it}$ is sufficiently weak to be ignored in computing permanent income. We estimate the parameters of equation (3) from race- and gender-specific mean regressions using all observations in which the person was either a head or wife. Our measure of permanent income is the individual-specific effect $v_i$, estimated as the mean of the residuals from the regression for each person. This measure is basically a time-average of past, current, and future income adjusted for demographic variables and time. We dropped individuals with less than 4 observations from the subsequent analysis to ensure that transitory income and measurement error will have only a minor effect on our permanent income measures.

**B Treatment of Outliers**

We eliminated extreme wealth values by first estimating separate median regression models for whites and blacks using the specification described in Section II above, pooling the residuals from those models, and dropping the observations corresponding to the bottom 0.5 percent and top 0.5 percent of the residuals. The procedure was conducted separately for couples, single men, and single women. We used separate trimming procedures for the level, ratio, and log models. Eliminating the outliers dramatically reduces the standard errors of our mean regression models estimates, at the likely cost of some bias. Our main findings are not sensitive to excluding the outliers but the size of the gap between whites and blacks is reduced somewhat. All results are for the trimmed sample unless noted otherwise.

**C Descriptive Statistics on Wealth, Income, and Demographics**

We begin by providing some basic facts about race differences in wealth and the key income and demographic variables. Table 1 provides the mean, median and standard deviation for the level of wealth ($W$), the wealth/permanent income ratio ($W/y$), and the log of wealth ($\ln W$). (Recall that, given the level of wealth $W$, the log of wealth is defined
as \ln \max\{W, 50\}. It also reports descriptive statistics for the levels and logs of current and permanent income and selected demographic variables. Outliers are included so that the numbers will better reflect population totals. The corresponding descriptive statistics for the trimmed sample can be found in Table A1 of Altonji and Doraszelski (2001) (hereafter, AD).

In the case of couples, the mean of wealth is 53,472 for blacks and 211,602 for whites, a ratio of 0.25. The race gap for current income is much smaller with a mean of 30,153 for blacks and 41,519 for whites, a ratio of 0.73. This is reflected in our permanent income measures, which have a mean of 32,450 for black heads and 46,949 for white heads. The permanent income values are 30,086 and 43,097 for black and white wives, respectively. The black/white ratio of permanent income is 0.70. Moreover, the distributions for current and permanent income are much more concentrated than the distributions for wealth. The much larger racial disparity in wealth is reflected in the wealth/permanent income ratios. The mean wealth/permanent income ratio for black couples is 1.80, which is only 37 percent of the value of 4.87 for white couples. The situations for single men and for single women mirror the one for couples. The means for many of the demographic variables differ considerably across races.

IV Decompositions of the Race Gap in Wealth

A Level Models

Below we present decompositions of the race gap in wealth into a component that is explained by differences in income and demographic variables and an unexplained component. The top row of Panel A in Table 2 reports decompositions of the wealth level $W_i$ for couples. The mean (standard error) of the wealth gap $\hat{W}_w - \hat{W}_b$ is 116,795 (4,535).\footnote{Using the coefficient estimates for the white equation $\hat{\theta}_w = (\hat{\alpha}_w, (\hat{\beta}_w)' )'$ we find that the race difference in income and demographics explains 92,589 (4,855) of the wealth gap, or 79 percent. We obtain strikingly different results with the black equation. Using}
\( \hat{\theta}^b = (\hat{\alpha}_b^0, (\hat{\alpha}_b')', (\hat{\beta}_b')')' \) we are only able to explain 29,009 (4,509) or 25 percent of the wealth gap.

Our results for single men (see second row of Panel A of Table 2) echo the findings for couples. Using the coefficient estimates for whites we find that black men would have 20 percent more wealth than white men if they had the same income and demographics as white men. This result suggests that the large wealth gap simply reflects racial differences in income streams, human capital variables, and current and past demographic variables. However, we again obtain very different results when we use the wealth model for blacks to perform the wealth decomposition. In this case only 31 percent of the gap of 40,365 (2,613) is attributable to income and demographics.

As the third row of Panel A of Table 2 shows, for single women the race gap in wealth is 46,575 (2,204). The model for the white sample implies that black women would have 103 percent of the wealth that white women hold if they had the same income and demographics as white women. However, the wealth model for blacks implies that only 33 percent of the gap is attributable to income and demographics.

The large discrepancy between the white and the black wealth models in the degree to which racial differences in income and demographic variables explain the wealth gap is a key theme in our analysis. It reflects the fact that wealth differences among blacks are much less sensitive to differences in income and demographics than wealth differences among whites.

To gain further insight into this key property of the wealth equations, we investigate the effect of a unit increase in current income \( y_{it} \) and permanent income \( y_i \) at the sample mean (pooled over race) of \( y_{it} \) and \( y_i \). The effect of a unit increase in current income is 0.64 for white couples and 0.69 for black couples. The effect of increasing the husband’s (wife’s) permanent income by one dollar is 1.04 (1.98) for whites and -0.02 (0.91) for blacks. The combined effect of increasing \( y_{it} \), \( y_i \) of the husband, and \( y_i \) of the wife by one dollar is 3.66 for whites and 1.58 for blacks. The point here is that these income derivatives tend to be much larger for white couples than for black couples. The same is true for single males and
single females. It also holds when we evaluate the income derivatives at 0.5 or 1.5 of the sample mean instead of at the sample mean. In fact, the gap in the sensitivity of wealth to income is substantially higher above the mean than below the mean.\footnote{Although there is some variation across variables (in part because of sampling error), the coefficients on the other variables in the model also tend to be larger in absolute value for whites than for blacks. To establish this point, we computed the indexes $Y_i \hat{\alpha}^w$ and $Y_i \hat{\alpha}^b$ of the income variables for each observation in the combined sample of blacks and whites. The regression of $Y_i \hat{\alpha}^b$ on $Y_i \hat{\alpha}^w$ and a constant is 0.284 with an OLS standard error of 0.0026. A similar regression involving the $X_i \hat{\beta}^w$ and $X_i \hat{\beta}^b$ indexes of the demographic variables shows that differences in demographics have a stronger association with wealth levels for whites than for blacks. The slope coefficient in a regression of $X_i \hat{\beta}^b$ on $X_i \hat{\beta}^w$ and a constant is 0.354 with an OLS standard error of 0.0016. When one forms the index using both income and demographic variables and regresses $Z_i \hat{\theta}^b$ on $Z_i \hat{\theta}^w$ and a constant, the slope coefficient is 0.305 (0.0023). The patterns are very similar for single men and single women, with the exception that for single men the coefficient on the index of income variables is close to 1.}

In summary, we find that most or all of the race gap in the wealth level for single men and single women and a substantial portion of the gap for couples would disappear if blacks and whites had the same distribution of income and demographic variables and the white wealth equation held for blacks. However, the wealth models for blacks exhibit much less sensitivity, indicating that both the race gap in income and demographics and race differences in the distribution of wealth conditional on income and demographic variables play important roles in the gap in wealth levels.\footnote{In Panel B of Table 2 we present results pertaining to the ratio of wealth to permanent income $W_i/y_i$. For white couples, the mean of the predicted value of $W_i/y_i$ is 4.18 (0.08), while the corresponding value for black couples is 1.74 (0.10). The total gap is 2.44 (0.13).}

B Ratio Models

In Panel B of Table 2 we present results pertaining to the ratio of wealth to permanent income $W_i/y_i$. For white couples, the mean of the predicted value of $W_i/y_i$ is 4.18 (0.08), while the corresponding value for black couples is 1.74 (0.10). The total gap is 2.44 (0.13).
Demographic and income characteristics account for 57 percent of this when we use white coefficients but only 12 percent when we use the black coefficients. In the case of single males the total gap is 0.70 (0.07). Demographic and income characteristics explain 119 percent of the gap when we use the white coefficients but only 27 percent when we use the black coefficients. In the case of single females, the corresponding numbers are 78 percent and 45 percent.

C Log Models

In Panel C of Table 2 we report decompositions for $\ln W_i$. The explanatory variables are the same as for the level and ratio models except that we substitute the log of current income and the log of permanent income for the levels of these variables. For couples the mean of the log wealth gap is 1.51 (0.06). Using the wealth model for whites we can explain 77 percent of this gap. Using the wealth model for blacks we can explain 72 percent. For single women, the mean of the gap in the log of wealth is 2.57 (0.08). Using the white equation the portion of the gap that is explained by income and demographics is 1.68 (0.10) or 65 percent of the total. The explained gap based on the equation for blacks is also 65 percent of the total gap. For single men the gap in log wealth is 2.00 (0.10). Using the white equation the portion of the gap that is explained by income and demographics is 1.71 (0.10), or 85 percent of the total. The explained gap based on the equation for blacks is 1.53 (0.12), or 76 percent of the total. In percentage terms, the explanatory variables account for more of the race gap in log wealth for single men than for either single women or couples. The 85 percent figure when the log wealth model for whites is used is particularly striking, and well in excess of the figures reported by Blau and Graham (1990).

The results based upon the white log model are less dramatic than the results for the white level model, and the portion of the gap explained using the black model is much larger in case of the log model than in case of the level model. However, this disparity between the log results and the results for levels and ratios should not be overstated. In particular, since the model for $\ln W_i$ implies a multiplicative model for $W_i$, the large race
 gap in the mean of log wealth together with the substantial fraction of the gap that is not explained by income and demographics implies that the response of wealth to income and demographic variables is much smaller for blacks than for whites. This point comes through most starkly in the case of single females. From Panel C of Table 2 the total gap in the mean of $\ln W_i$ is 2.57. The regression models for whites and blacks both explain 65 percent of this gap, leaving an unexplained gap of 0.90 ($= 0.35 \times 2.57$). Consider the most extreme case in which the slope coefficients of the log models for whites and blacks are equal and all of the unexplained gap is due to the intercept. Assume that the distribution of the errors in the log models do not depend on race, income, and demographics. Then for whites the derivative of wealth with respect the income and demographics will be $e^{-0.90} \approx 2.46$ times the corresponding derivative for blacks. Consequently, the evidence based upon the log models is broadly consistent with the evidence for levels and ratios.

D Self Employment and the Wealth Gap

As can be seen from Table 1, there is a considerable race gap in the self-employment rate (0.18 versus 0.05 in the case of heads and 0.08 versus 0.03 in the case of spouses). This confirms a large literature showing that the black self-employment rate is only about one third of the white self-employment rate and that this ratio has been relatively constant for the past 70 years (Fairlie and Meyer 1997).

If causality runs mainly from self employment to wealth, then it is desirable to control for self employment in the analysis, especially to the extent that the effects of past discrimination on the self-employment rate of blacks lingers today (see for example the discussion in Bates (1997) and Oliver and Shapiro (1995)). However, some studies of the race gap in self employment attribute part of the self-employment gap to a lack of financial capital. If the causality runs from wealth to self employment, then it is less clear that self employment should be controlled for.

The coefficient on self employment is quite substantial. In the case of white couples, the coefficients on the dummies for self employment of the head and self employment of the
spouse are 150,010 and 47,916 respectively in the model for wealth levels. The corresponding coefficients in the model for blacks are 57,415 and -3,784. The combination of a much higher self-employment rate for whites and a much stronger association between self employment and wealth for whites make a substantial contribution to the race gap in wealth. When we remove the self-employment indicators from the wealth model for couples, the fraction of the gap in $W_i$ explained using the white wealth model declines from 79 percent to 64 percent. For $W_i/y_i$ the decline is from 57 percent to 41 percent. Differences in self-employment rates make a much smaller contribution to the wealth gap for single males and single females. The percentage of the gap in $W_i$ explained using the white wealth model is 114 percent for males and 100 percent for females even when self employment is excluded. Overall, the evidence suggests that race differences in self-employment patterns are a significant part of a full explanation of black/white wealth differences, although causality is ambiguous.

E Alternative Samples and Evaluation Points for the Wealth Gap

In this section we explore the possibility that a lack of overlap between the white and black samples is leading to unreliable estimates, particularly when the black coefficients are used with white characteristics, because we are extrapolating too far out of sample.

First, when we decompose the gap at the sample means of the explanatory variables for whites and the sample means for blacks we obtain results similar to those based on the full distribution for each group. This suggests that the right tail of the distribution of current income and/or permanent income is not driving our results.

Second, when we add third powers in current and permanent income to address the possibility that the effects of current and permanent income on wealth is not quadratic over the full range of the distribution of whites and blacks, our decompositions based on the white wealth model are not affected very much. However, the decompositions based on the black wealth model become very imprecise in the case of single women. For this group
there is not enough overlap in the income distributions at the high end.

As a final robustness check we start with the trimmed sample and prior to estimation of the wealth models eliminate observations if current income is negative or if permanent income is less than 100. We also eliminate observations for whites with current or permanent income above the corresponding 98th percentile values of the black sample. In the case of couples, we screen on the basis of both the husband’s permanent income and the wife’s permanent income. We include cubics in both current income and permanent income. The gap in $W_i$ is 85,644, which is below the gap of 116,795 in the full sample because we have chosen whites who fall more within the income distribution of blacks. Income and demographics explain 64 percent of the gap between whites and blacks using the white wealth model and 26 percent using the black wealth model. The respective figures are 135 percent and 19 percent for single males and 97 percent and 28 percent for single females. The respective figures for $W_i/y_i$ are 50 percent and 19 percent for couples, 142 percent and 25 percent for single males, and 61 percent and 36 percent for single females. We obtain very similar results using a quadratic specification. Overall, the results are quite similar to those in Table 2, especially in view of the fact that eliminating high income whites reduces the potential size of the explained gap.

Barsky et al. (2002) take a nonparametric approach to addressing the potential for bias from misspecification of the relationship between wealth and income in the presence of differences in the white and black income distributions. They adjust for differences in income by reweighting the observations on whites to have the same income distribution as the black sample and find that differences in income account for 64 percent of the wealth gap. Regarding the question whether differences in characteristics explain less of the gap when the black wealth function is used, Barsky et al. (2002) argue that there are not enough blacks at the high end of the white income distribution. Our sample is much larger than theirs. Moreover, our robustness checks above suggest strongly that in our samples overlap in the distributions is sufficient to permit decompositions using the black wealth model, particularly if one restricts the range of income variation in the white sample. The
discrepancy between the white and black wealth models in the importance of income and
demographic variables appear to be real rather than an artifact caused by inaccurate
extrapolation out of sample. However, further research on this key issue is needed.

F  Summary

Our results show substantial differences in the sensitivity of wealth holding to income and
demographic variables. Because both income and demographic characteristics of whites are
more conducive to wealth holding, we assign much higher fractions of the wealth gap to
differences in income and demographics when we use the white wealth equations. Most or
all of the race gap in the wealth level for single men and single women and a substantial
portion of the gap for couples would disappear if blacks and whites had the same
distribution of income and demographic variables and the white wealth equation held for
blacks. Although the relative explanatory power of the regression models for blacks is
higher when $W_i/y_i$ is the explanatory variable than when $W_i$ is the explanatory variable,
the results for $W_i/y_i$ are generally consistent with the results for $W_i$. Finally, the evidence
based upon the log models is broadly consistent with the evidence for levels and ratios.

The median regression estimates parallel the mean regression estimates, but the
percentages of the gap explained using the white wealth model are somewhat lower.
Details can be found in Appendix 1. Similarly, results for main home equity (house value
net of mortgage balance), wealth in farms/businesses, and stocks/mutual funds and IRAs
closely parallel the results for total wealth. For example, income and demographics explain
88 percent of the gap in main home equity when we use the white coefficients but only 30
percent when we use the black coefficients. The details are in AD. In the next two sections,
we consider possible explanations for why wealth holdings are more sensitive to income and
demographics for whites than for blacks.
V Intergenerational Transfers and the Wealth Gap: Evidence for Siblings

There are at least three possible explanations for why wealth holding are more sensitive to characteristics for whites than for blacks. First, whites may enjoy a higher rate of return on assets, in which case the same level of savings and intergenerational transfers would lead to larger wealth levels, magnifying underlying differences that are associated with income and demographics. Menchik and Jianakoplos (1997) provide some limited evidence that blacks experience a lower rate of return on assets of a given type. However, the evidence on this point is far from conclusive (see also Gittleman and Wolff (2001)). Second, savings of blacks may be less sensitive to income and demographic variables. Third, inheritances of housing, financial assets and businesses and/or inter vivos transfers are larger among whites than among blacks because slavery and the legacy of racial discrimination have inhibited the accumulation of wealth in the black population. Studies by Smith (1995), Avery and Rendall (1997), Menchik and Jianakoplos (1997), and Gittleman and Wolff (2001) indeed suggest that differences in inheritances explain part of the race gap, although none of these studies pays much attention to the fact that there are large discrepancies between whites and blacks in the sensitivity of wealth holding to income and demographics.

In this section we use sibling models with family fixed effects to explore the possibility that differences in intergenerational transfers are the source of differences in wealth holding. Using siblings largely neutralizes the effects of differences between whites and blacks in inter vivos transfers and inheritances and provides a way of controlling for the effects of adverse history on the position of blacks relative to whites with similar income and demographics.
A Family Fixed Effects as a Control for Intergenerational Transfers

Let \( k \) index families and \( i \) siblings within a family. We estimate models of the form

\[
W_{kit}^j = \alpha^j_0 + Y_{kit}^j \alpha^j + X_{kit}^j \beta^j + u_k^j + \epsilon_{kit}^j, \quad j = b, w,
\]

where \( u_k^j \) is a family-specific fixed effect. For reasons that will become clear later, we have made the wealth survey year subscript \( t \) explicit. If inheritances and inter vivos transfers are correlated with average income and demographic characteristics of the siblings, then \( u_k^j \) will be correlated with \( Y_{kit}^j \) and \( X_{kit}^j \). Since inheritances and transfers are larger for whites than blacks, the resulting bias may be larger for whites. By controlling for factors that are common to siblings, we reduce this problem.\(^{11}\)

To see how historical barriers to wealth holding faced by blacks affect the estimated slopes of wealth models and how fixed effects models reduce the influence of these barriers, consider the following simple model of inheritances and inter vivos transfers. Suppose that in a steady state the wealth of parents, including inheritances and accumulated savings, is related to permanent income \( y \) according to

\[
W_0 = \bar{W}_w + \alpha y_0, \quad \alpha > 0,
\]

where \( \bar{W}_w \) is a constant for whites and we have suppressed family and survey year subscripts and use the subscript 0 to denote the parents. Suppose also that parents pass on the fraction \( \phi \) of their wealth. Furthermore, assume that the relationship between the permanent incomes of parents and children is

\[
y_0 = \bar{y}_w + \rho y_1, \quad \rho > 0,
\]

where \( \bar{y}_w \) is a constant for whites and the subscript 1 denotes the child. For whites, the
child’s wealth is the sum of savings out of gifts and inheritances and savings out of income,

$$W_1 = s(\phi W_0 + gy_1) = s\phi(\bar{W}_w + \alpha \bar{y}_w) + (s\alpha \phi \rho + sg)y_1$$

where $g$ converts permanent income into the discounted sum of lifetime income and $s$ is the savings rate. Suppose that slavery and discrimination severely limited the ability of blacks to accumulate wealth. Then the relationship between wealth and permanent income of the black parents can be modelled as

$$W_0 = \tilde{W}_b + d(\alpha y_0), \quad 0 < d < 1,$$

where $d$ reflects the effect of discrimination. Hence, the equation relating wealth to income for blacks is

$$W_1 = s(\phi W_0 + gy_1) = s\phi(\tilde{W}_b + d\alpha \bar{y}_b) + (s\alpha \phi \rho d + sg)y_1$$

where $\bar{y}_b$ is the intercept in the equation $y_0 = \bar{y}_b + \rho y_1$ relating $y_0$ to $y_1$ for blacks. The coefficient $(s\alpha \phi \rho d + sg)$ on $y_1$ in the above equation for blacks is smaller than the coefficient $(s\alpha \phi \rho + sg)$ for whites. However, once the effect of parental wealth is held constant using a family-specific fixed effect, the influence of the term $s\phi W_0$ is eliminated and the coefficient on $y_1$ is $sg$ for both whites and blacks. A similar argument can be made for intergenerational links in demographic patterns. Hence, if the analyses with and without fixed effects give similar answers, then we can conclude that race differences in gifts and inheritances that are correlated with income and demographic variables do not explain our finding that wealth levels are more sensitive to income and demographic variables in the case of whites than in the case of blacks.

**B Results**

To obtain adequate sample sizes, we pool observations on single men, single women, and couples and add control variables for the three demographic groups. For couples we use
only the head’s variables rather than separate variables for the husband and the wife. The
decompositions are unweighted.

When one includes a family fixed effect (FE), the effects of permanent income are
identified only by cross-sibling variation, but the effects of current income, self-employment
status, marital status, the number of children, and other explanatory variables that change
over time are identified by a mix of cross-sibling variation and cross-time variation. OLS
without fixed effects uses cross-family, cross-sibling, and cross-time variation. One might
argue that cross-person variation should have a stronger relationship with wealth than
cross-time variation. Moreover, for the purpose of explaining the race gap in wealth the
effects of permanent differences are more important than transitory differences. For this
reason, we also examine estimates that emphasize cross-sibling variation. A simple way to
use only cross-sibling variation in the key income and demographic variables is to estimate
the wealth model after taking person means across wealth surveys for all of the variables,
and we refer to the resulting estimates as FE-Means and OLS-Means. However, with time
averaging it becomes difficult to distinguish the effects of current and permanent income,
and much variation in variables such as the survey year dummies and age is lost.

To save space, we do not report the detailed regression estimates. However, it is
interesting to note that although the race gap in the effect of income on wealth is a little
smaller for FE than OLS it is somewhat larger for FE-Means than OLS. Overall, there is
not much evidence that race differences in the correlation between income and inheritances
plays a large role in the stronger relationship between income and wealth for whites.

There is an interesting pattern in coefficients on self employment. The OLS and FE
estimates of the effect of self employment are 70,013 and 49,116 for whites. The
OLS-Means and FE-Means estimates are 85,232 and 60,751. Both reporting error and the
fact that successful businesses are longer lasting may underlie the fact that the estimates
are larger when we work with individual means. The fact that the estimates decline when
sibling fixed effects are included is consistent with a role for inheritances in the effects of
self employment but it may also be due to a greater role for reporting error. The OLS and
FE estimates for blacks are 14,230 and 11,854 and the OLS-Means and FE-Means estimates are 12,116 and 9,804. Thus, adding family fixed effects closes the race gap in the link between self employment and wealth from 73,116 to 50,948 when individual means are taken and from 55,783 to 37,262 when means are not taken. Overall, the results for siblings confirm our earlier results suggesting that self employment may play a substantial role in the wealth gap. The comparison of the estimates with and without fixed effects suggests that parental wealth or other family background factors can explain about one third of the large race gap in the relationship between self employment and wealth.\textsuperscript{13}

In Table 3 we report decompositions of the wealth gap using the siblings sample using FE and FE-Means estimates of the wealth models. For purposes of comparison, we also report OLS and OLS-Means estimates. The results are quite striking. For the wealth level we explain 111 percent and 54 percent of the gap using the white coefficients and the black coefficients, respectively, using OLS, 104 percent and 49 percent using FE, 107 percent and 36 percent using OLS-Means, and 97 percent and 23 percent using FE-Means. Basically, the FE and FE-Means results closely correspond to the results without fixed effects, and mirror the pattern we obtained above for the samples of couples, single females, and single males. We continue to explain more of the wealth gap using the white coefficients than the black coefficients, particularly when wealth and income are specified in levels. The $W/Y$ results are very similar to the $W$ results. If anything, the gap in the fraction of wealth explained using the black wealth model and the white wealth model is larger when we use cross-sibling variation to eliminate the effects of common inheritances and transfers. The fixed effects results for $\ln W$ are also similar to our earlier results in Table 2. In sum, there is little evidence that differences in factors such as inheritances or inter vivos transfers that are likely to vary across families provide an explanation for the racial difference in the sensitivity of wealth to income and demographics. We wish to stress, however, that they may play a role in differences in wealth intercepts and appear to matter for self employment. This leaves differences in savings behavior and/or rates of return as potential sources of the race difference in wealth models. We examine them in the next section.
There is an important caveat to our analysis. The additively separable form of equation (4) is standard in the literature, but may not be adequate if income and inheritances interact in the wealth function, as would be the case if there is a strong bequest motive. The results for $W/y$ and $\ln W$ are less subject to this objection.

VI Decompositions of the Race Gap in the Growth of Wealth

We now turn to wealth accumulation, which we measure using the 5 year changes between the 1984 and 1989 surveys or between the 1989 and 1994 surveys. These show a much more rapid rate of accumulation for whites than for blacks. In the trimmed sample, for example, the mean 5 year change in wealth including home equity is 36,160 for whites and 8,237 for blacks. The corresponding values for $W/y$ are 1.28 and 0.45, respectively. The large difference could reflect differences in savings, differences in rates of return on assets, or a combination of the two.

How much of the difference in accumulation rates can be explained by differences in income and demographics? To answer this question, we estimate growth models that include the variables that appear in the wealth models plus the changes in the measures of current income, region, SMSA, wife’s work hours, children, dependents, health status, and self employment. To insure reasonable sample sizes we confine the analysis to couples. We do not weight observations in computing the decompositions.

The wealth change decompositions are reported in Table 4. In the case of $W$, income and demographic variables explain 74 percent of the difference in accumulation rates when the accumulation model for whites is used. Using the accumulation model for blacks, the same factors explain 49 percent of the gap. The corresponding figures for growth in the wealth/permanent income ratio, $W/y$, is 84 percent based on the white accumulation model and 30 percent based on the black accumulation model. In summary, we find that income and demographic characteristics explained more of the gap in wealth accumulation.
when the wealth model for whites is used.

What is responsible for the substantial difference between blacks and whites in the sensitivity of wealth levels and wealth/permanent income ratios to income and demographic variables? Difference in the rate of return to wealth may be part of the story, because blacks have larger percentages of their wealth in home equity and smaller percentages in stocks and business wealth. However, the differences seem too great to be attributable to differences in rate of return alone although further research on this issue is needed. The wealth change regressions suggest that differences in savings rates are an important factor. Recall that our sibling results show that inheritances and gifts are not a major factor in the black/white differences in the wealth models. This fact, in combination with the substantial race differences in the wealth growth models, suggests that differences in savings behavior, possibly in combination with rates of return, may be an important source of the wealth gap and of black/white differences in the wealth models.\textsuperscript{14}

Note that the results for logs in Panel C of Table 4 present a challenge to the above interpretation. The results for logs show that 77 percent of the wealth is explained when the black wealth growth model is used while only 5 percent of the gap is explained when the white log wealth growth model is used. These results are not consistent with the results in Panels A and B and cannot be attributed to the fact that the log model is basically a multiplicative specification. We do not have a good explanation for this, but note that many of the coefficients in the model are poorly estimated and the explanatory power of both the white model and black model is low. In general, the low explanatory power of the growth models makes us cautious in interpreting the decompositions, especially in the case of the models for blacks.\textsuperscript{15}

In independent work, Gittleman and Wolff (2001) analyze race differences in wealth accumulation using a different approach than ours. They find that race differences in savings rates are entirely explained by differences in income and demographic differences. Because they pool whites and blacks and use population weights, their regression estimates should be close to those for the white sample. Thus, their results for savings are
qualitatively consistent with our results for the change $W/y$, although we only explain 84 percent of the gap in the change in $W/y$.

VII Conclusions and Further Questions

Using better income and demographic controls than were available for previous studies, we can explain a large part of the racial disparity in wealth holdings with income and demographic variables provided that we estimate the wealth model on a sample of whites, but only a small fraction when we use the black wealth model. We also find that a higher self-employment rate and a stronger link between self employment and wealth for whites than for blacks make an important contribution to the wealth gap between white and black couples. Overall, our results suggest that race differences in the sensitivity of wealth to income and demographics may be as important as the gap in income and demographics in understanding black/white differences in wealth.

What underlies the substantial race differences in the sensitivity of wealth holding to income and demographics? The fact that we obtain similar results when we relate sibling differences in wealth to sibling differences in income and demographics suggests, perhaps surprisingly, that differences in inter vivos transfers and inheritances are not the main reason the wealth model coefficients differ by race. This would seem to leave race differences in savings behavior and/or rates of return as a default explanation. We find that income and demographic differences explain a substantial part of the difference between whites and blacks in the growth in the level of wealth and in the ratio of wealth to permanent income when we use the wealth model for whites, but relatively little when we use the wealth model for blacks. This result points to important differences between whites and blacks in the effects of income and demographics on savings and/or rates of return, although this conclusion is tentative for a number of reasons discussed above. We suspect that differences in savings behavior is the main factor. This conjecture is supported by the fact that about 25-30 percent of black households are “unbanked,” meaning that they have no direct access to a financial institution (Hogarth and O’Donnell 1999, Hogarth and
Lee 2000).

A large research agenda remains. Our sibling approach to the study of the effects of parental resources on the wealth models should be augmented by a study of the sources of intergenerational links in wealth along the lines of recent work by Charles and Hurst (2002). The role of self employment and small business formation has already received a substantial amount of attention in the literature and deserves more.

As suggested by Smith (1995), a possible explanation for the racial disparity in savings behavior is that differences in permanent income in conjunction with lower life expectancies among blacks and a higher replacement rate of private income with publicly provided social security and health benefits depress the incentive to save for retirement of blacks relative to whites.\textsuperscript{16} Lower life expectancy for blacks should reduce savings, although we doubt that it plays a major role.

The lower incomes of black households may lead to greater dependence on social transfer programs relative to own resources and private transfers, holding permanent income and other household characteristics constant. Eligibility requirements for many transfer programs, including welfare programs and public housing, discourage wealth accumulation. This may reduce the private incentive to accumulate wealth.\textsuperscript{17}

Another possible explanation revolves around the implications of differences in the income distributions of the friends and relatives of whites versus blacks. The basic idea is that savings and wealth accumulation are influenced by economic links to other households and friends that are motivated by altruism or other factors. The effects of ties to other households on savings depends on the level and distribution of resources and needs. We know that wealth is a convex function of permanent income. An explanation for this is that the marginal utility of household consumption declines in spending, leading households to accumulate resources as a hedge against future consumption needs, fluctuations in income, and uncertainty about the lifespan. Unspent resources are left to children or to charity late in life or upon death. However, if a high income family has strong ties to needy relatives and friends, it may make transfers rather than accumulate wealth. Also, social pressure on
the relatively well off to provide assistance may be stronger in communities where needs are greatest. This will reduce the private incentive to accumulate for those who are not very altruistic. The net result may be that the wealth of a household with an extensive network of needy family members and friends may increase less with income than the wealth of a household with a network of well off friends and relatives.

Appendix 1  Median Regression Results

There are a number of reasons why median regression estimates are of interest. First, one may simply be more interested in the median of wealth than in the mean of wealth for persons with a given set of characteristics. Second, the skewness and fat tails of the wealth distribution suggest that it may be easier to estimate conditional medians (Koenker and Bassett 1978, Narula and Wellington 1982).

In light of space constraints, we report only selected decompositions. In the case of couples, the total gap in the conditional median of wealth, $\bar{W}_w - \bar{W}_b$, is estimated to be 84,794 (2,824), which compares to a gap in the mean of wealth of 116,795 (4,535). The gap in $W/y$ is 1.81 (0.07) for medians compared to 2.44 (0.13) for means. The gaps in the median of $\ln W$ and the mean of $\ln W$ are more similar (1.36 (0.05) versus 1.51 (0.06)), which is not surprising given that the log transformation reduces the skewness of the wealth distribution.

In the case of couples the income and demographic variables account for 68 percent of the wealth gap in the weighted average of the conditional medians if the median regression function for whites is used. In contrast, the median wealth regression for blacks implies that characteristics account for only 22 percent of the gap in the conditional median of wealth. For single men the white median regression implies that 85 percent of the gap is explained, while the black median regression implies a figure of only 41 percent. The corresponding figures for single women are 74 percent and 42 percent. Overall, income and demographics account for a somewhat smaller percentage of the race gap in the conditional median of the wealth level than in the conditional mean when the white model is used,
particularly in the case of single men and single women. When the median wealth model for blacks is used, the percentage explained is slightly lower than the corresponding value for the conditional mean in the case of couples but higher for single men and women.

The results for $W/y$ are similar to the results for levels, except that the portion of the gap explained by the white median regression model is substantially smaller for both couples and single females. The results for $\ln W$ are basically similar to the results based on mean regression and also imply that the wealth of whites is much more sensitive to income and demographics than the wealth of blacks once the implications of the log specification and the large unexplained gap are taken into account.
References

[1]


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Barsky, R., Bound, J., Charles, K. K. and Lupton, J.


   Federal Reserve Bank, Washington.


Notes

1Smith (1995) finds that “healthier households are wealthier ones” for both blacks and whites. Hence, controlling for health status helps to explain the wealth gap. The question of causality, however, is tricky. Similarly, we include the wife’s work hours although there are some obvious endogeneity issues that may lead to different biases for whites and blacks. The wealth decompositions in Table 2 below are not very sensitive to dropping it.

2An alternative measure of permanent income is based upon nominal non-asset family income prior to the year of the wealth survey. We typically explain as much or more of the wealth gap with this backward-looking measure than with our usual measure.

3Because we estimate the model parameters without weighting but weight the observations when performing decompositions, $\hat{W}_w - \hat{W}_b$ need not equal the difference in the weighted sample means of wealth. The latter figure is 115,228(=187,589-52,361).

4The effect of a unit increase in current and permanent income is 2.04 for whites and 2.04 for blacks at 0.5 of the pooled sample mean, which is the only exception to the rule of larger marginal effects for whites. At 1.5 of the mean the values are 5.28 for white couples versus 1.12 for black couples. The values for single white males and single black males are 1.00 versus 0.46 at 0.5 of the mean, 1.55 versus 0.39 at the mean, and 2.11 versus 0.31 at 1.5 times the mean. The values for single white females and single black females are 1.95 versus 0.82 at 0.5 of the mean, 1.75 versus 0.68 at the mean, and 1.55 versus 0.54 at 1.5 times the mean.

5Hurst, Luoh and Stafford (1998) use the PSID to examine race differences in the level of wealth. Their results are broadly consistent with ours. However, Hurst et al. (1998) do not directly address the question of differences in the form of the wealth function because they pool whites and blacks and only allow the intercept of the wealth model to depend on race. See footnotes 1 and 17 of AD for details.

6If $\ln W_i = Z_i \theta + \epsilon_i$ and $\ln W_i = .90 + Z_i \theta + \epsilon_i$, then $E(W_i | Z_i) = e^{Z_i \theta} E(e^{\epsilon_i})$ and $E(W_i | Z_i) = e^{.90 + Z_i \theta} E(e^{\epsilon_i})$.

7The corresponding values with and without self employment when the black wealth models are used are 25 percent and 19 percent for $W_i$ and 12 percent and 5 percent for $W_i/y_i$.

8The distributions of the head’s permanent income (in thousands) in the restricted samples are as follows:
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<th>min</th>
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<td>48</td>
<td>69</td>
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</table>

9 See Oliver and Shapiro (1995) for a historical overview of legal and social barriers to wealth holding by blacks.

10 If blacks face higher prices than whites, our measure of real income is overstated for blacks. The available evidence suggests that consumer prices are not very different for blacks than for whites (Richburg Hayes 2000).

11 The coefficients in the sibling equations will be downward biased to the extent that parents try to compensate for differences in income of children. However, Menchik (1980), Menchik and David (1983), Wilhelm (1996) and other studies show that inheritances are evenly split in about 70 percent of the cases and that the division is only weakly related to income in the cases where the split is not even. Inter vivos transfers among siblings tend to compensate for income differences, but the coefficient relating transfer amounts to income differences is relatively small. See McGarry and Schoeni (1995) and Altonji, Hayashi and Kotlikoff (1997) for evidence.

12 The fixed effect estimate of the marginal effect of a one dollar increase in both current income and permanent income is 1.49, 2.26, and 2.82 when both are set to 0.5, 1.0, and 1.5 times their respective means for the pooled sibling sample. The corresponding fixed effects estimates for blacks are 0.78, 1.62 and 2.46. The corresponding FE-Means estimates are 1.30, 2.02 and 2.73 for whites and 0.98, 0.99 and 0.99 for blacks. The corresponding OLS estimates are 1.77, 2.32 and 2.88 using the white model and 0.89, 1.52 and 2.15 using the black model.

13 Fairlie and Meyer (2000) consider estimates of intergenerational links in self employment and conclude that most of the racial discrepancy in self-employment rates is due to “forces that reduce current black self employment besides the initial conditions of low black self employment.” Dunn and Holtz-Eakin (2000) find that parental wealth has a small direct effect on self employment of sons.

14 We have also estimated equations for $W$ that include a lagged value of $W$ along with the other variables used in the savings regressions. However, given the presence of lagged wealth, the high explanatory power of the models is largely due to the race gap in lagged wealth, and so it is hard to know what the economic significance of the explained gap is. See footnote 29 of AD for details.

15 The adjusted $R^2$’s of the growth models are only about 0.05 for the growth in ln$W$. The adjusted $R^2$
is 0.0946 for whites and 0.0458 for blacks when the dependent variable is the growth in $W/y$ and 0.3023 for whites but only 0.0607 for blacks when the dependent variable is the growth in $W$.

^{16}In preliminary work using the Health and Retirement Survey (HRS), John Karl Scholz (University of Wisconsin, Madison) has found that the race gap is much smaller for pensions than other forms of wealth (personal communication). This could have spillovers into holding of assets.

^{17}See Powers (1998), Gruber and Yelowitz (1999), and Hurst and Ziliak (2001) for direct empirical evidence on the effects of asset limits on savings. The evidence is mixed. Hubbard, Skinner and Zeldes’s (1995) simulation analysis suggests asset limits may be important. Their simulation methodology could be adapted to address the other possibilities mentioned above.
## Table 1
Descriptive statistics for wealth and income variables

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<td>(standard deviation)</td>
<td>(standard deviation)</td>
<td>(standard deviation)</td>
</tr>
<tr>
<td>Log of wealth including main home equity</td>
<td>10.97 (2.11)</td>
<td>9.48 (2.35)</td>
<td>8.95 (2.76)</td>
<td>7.25 (2.82)</td>
<td>7.78 (2.80)</td>
<td>6.88 (2.88)</td>
</tr>
<tr>
<td>Wealth including main home equity</td>
<td>211602 (627938)</td>
<td>53472 (140253)</td>
<td>62211 (153775)</td>
<td>15439 (38141)</td>
<td>2387 (208029)</td>
<td>70599 (40543)</td>
</tr>
<tr>
<td>Ratio of wealth including main home equity to permanent income</td>
<td>4.87 (11.26)</td>
<td>1.80 (4.36)</td>
<td>1.33 (3.12)</td>
<td>0.58 (1.68)</td>
<td>0.07 (7.50)</td>
<td>2.04 (1.88)</td>
</tr>
<tr>
<td>Log of taxable non-asset income</td>
<td>10.35 (0.80)</td>
<td>10.06 (0.78)</td>
<td>9.68 (0.90)</td>
<td>9.12 (1.07)</td>
<td>9.33 (0.88)</td>
<td>9.27 (0.90)</td>
</tr>
<tr>
<td>Permanent log-income</td>
<td>10.72 (0.49)</td>
<td>10.32 (0.54)</td>
<td>10.56 (0.57)</td>
<td>9.91 (0.72)</td>
<td>10.52 (0.56)</td>
<td>9.57 (0.63)</td>
</tr>
<tr>
<td>Total taxable non-asset income</td>
<td>41519 (43788)</td>
<td>30153 (19985)</td>
<td>22452 (19181)</td>
<td>14204 (11887)</td>
<td>11283 (26781)</td>
<td>14618 (8844)</td>
</tr>
<tr>
<td>Permanent income</td>
<td>46949 (20613)</td>
<td>32450 (12747)</td>
<td>45317 (17643)</td>
<td>28413 (9683)</td>
<td>27217 (13731)</td>
<td>38939 (7872)</td>
</tr>
<tr>
<td>Spouse: Permanent log-income</td>
<td>10.66 (0.49)</td>
<td>10.25 (0.56)</td>
<td>10.37 (0.56)</td>
<td>10.25 (0.56)</td>
<td>10.37 (0.56)</td>
<td>10.25 (0.56)</td>
</tr>
<tr>
<td>Spouse: Permanent income</td>
<td>40397 (22638)</td>
<td>30086 (13881)</td>
<td>43148 (12368)</td>
<td>30770 (13881)</td>
<td>27970 (13881)</td>
<td>27970 (13881)</td>
</tr>
<tr>
<td>Head: Health: fair or poor</td>
<td>0.14 (0.29)</td>
<td>0.16 (0.29)</td>
<td>0.16 (0.29)</td>
<td>0.26 (0.29)</td>
<td>0.25 (0.29)</td>
<td>0.25 (0.29)</td>
</tr>
<tr>
<td>Head: Education: grade school</td>
<td>0.08 (0.18)</td>
<td>0.18 (0.27)</td>
<td>0.07 (0.27)</td>
<td>0.12 (0.27)</td>
<td>0.13 (0.27)</td>
<td>0.13 (0.27)</td>
</tr>
<tr>
<td>Head: Education: high school incomplete</td>
<td>0.13 (0.19)</td>
<td>0.14 (0.20)</td>
<td>0.14 (0.20)</td>
<td>0.20 (0.20)</td>
<td>0.16 (0.20)</td>
<td>0.28 (0.20)</td>
</tr>
<tr>
<td>Head: Education: high school diploma</td>
<td>0.24 (0.28)</td>
<td>0.22 (0.29)</td>
<td>0.22 (0.29)</td>
<td>0.29 (0.29)</td>
<td>0.28 (0.29)</td>
<td>0.28 (0.29)</td>
</tr>
<tr>
<td>Head: Education: high school diploma plus</td>
<td>0.29 (0.24)</td>
<td>0.32 (0.30)</td>
<td>0.32 (0.30)</td>
<td>0.30 (0.30)</td>
<td>0.26 (0.30)</td>
<td>0.26 (0.30)</td>
</tr>
<tr>
<td>Head: Education: college degree</td>
<td>0.17 (0.06)</td>
<td>0.17 (0.08)</td>
<td>0.17 (0.08)</td>
<td>0.20 (0.08)</td>
<td>0.11 (0.08)</td>
<td>0.28 (0.08)</td>
</tr>
<tr>
<td>Head: Education: advanced or professional degree</td>
<td>0.10 (0.05)</td>
<td>0.08 (0.01)</td>
<td>0.08 (0.01)</td>
<td>0.08 (0.01)</td>
<td>0.06 (0.01)</td>
<td>0.06 (0.01)</td>
</tr>
<tr>
<td>Head: Self-employed</td>
<td>0.18 (0.05)</td>
<td>0.10 (0.05)</td>
<td>0.10 (0.05)</td>
<td>0.05 (0.05)</td>
<td>0.05 (0.05)</td>
<td>0.05 (0.05)</td>
</tr>
<tr>
<td>Head: Number of siblings</td>
<td>2.30 (2.65)</td>
<td>3.73 (3.96)</td>
<td>3.00 (3.96)</td>
<td>2.00 (3.96)</td>
<td>2.00 (3.96)</td>
<td>2.00 (3.96)</td>
</tr>
</tbody>
</table>

Number of observations:
- from 1984 wealth survey: 7700 (2509) 1415 (1130) 2744 (3178)
- from 1989 wealth survey: 2561 (910) 442 (335) 898 (1036)
- from 1994 wealth survey: 2777 (931) 502 (426) 912 (1090)
- Number of couples/singles: 3444 (1280) 924 (701) 1562 (1587)

Notes: Computed from the pooled sample using weights. The weights are normalized so that for each subgroup the means are estimates of the average of the population means for 1984, 1989, and 1994. We show the mean (left), median (right), and standard deviation for each variable by demographic group. The definition of permanent income is given in the text.
Table 2  
Mean regression decompositions of the race gap in wealth

<table>
<thead>
<tr>
<th>Demographic group</th>
<th>White characteristics (standard error)</th>
<th>Black characteristics (standard error)</th>
<th>Black characteristics (standard error)</th>
<th>White characteristics (standard error)</th>
<th>Total gap (standard error)</th>
<th>Explained gap using white coefficients (standard error) (percent)</th>
<th>Explained gap using black coefficients (standard error) (percent)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
<td>(5)</td>
<td>(6)</td>
<td>(7)</td>
</tr>
<tr>
<td>couples</td>
<td>166708</td>
<td>74119</td>
<td>49912</td>
<td>78921</td>
<td>116795</td>
<td>92589</td>
<td>29009</td>
</tr>
<tr>
<td></td>
<td>(3344)</td>
<td>(4280)</td>
<td>(3064)</td>
<td>(6335)</td>
<td>(4535)</td>
<td>(4855)</td>
<td>(4509)</td>
</tr>
<tr>
<td></td>
<td>(79%)</td>
<td>(25%)</td>
<td>(65%)</td>
<td>(28%)</td>
<td>(20%)</td>
<td>(87%)</td>
<td>(26%)</td>
</tr>
<tr>
<td>single males</td>
<td>54164</td>
<td>5763</td>
<td>13800</td>
<td>26205</td>
<td>40365</td>
<td>48402</td>
<td>12405</td>
</tr>
<tr>
<td></td>
<td>(2421)</td>
<td>(4787)</td>
<td>(982)</td>
<td>(2281)</td>
<td>(2613)</td>
<td>(5435)</td>
<td>(2154)</td>
</tr>
<tr>
<td></td>
<td>(120%)</td>
<td>(31%)</td>
<td>(110%)</td>
<td>(32%)</td>
<td>(13%)</td>
<td>(122%)</td>
<td>(137%)</td>
</tr>
<tr>
<td>single females</td>
<td>59268</td>
<td>11119</td>
<td>12693</td>
<td>28260</td>
<td>46575</td>
<td>48149</td>
<td>15567</td>
</tr>
<tr>
<td></td>
<td>(2048)</td>
<td>(3896)</td>
<td>(816)</td>
<td>(4411)</td>
<td>(2204)</td>
<td>(4447)</td>
<td>(4163)</td>
</tr>
<tr>
<td></td>
<td>(103%)</td>
<td>(33%)</td>
<td>(33%)</td>
<td>(33%)</td>
<td>(33%)</td>
<td>(33%)</td>
<td>(33%)</td>
</tr>
</tbody>
</table>

Notes: Computed from the trimmed pooled samples using weights (see text for details on the trimming). The regression coefficient estimates are estimated from the trimmed pooled samples without sample weights (see Tables A2-A4 for mean regression results). Columns (1) and (2) are based on coefficient estimates from the white sample, columns (3) and (4) on coefficient estimates from the black sample. Column (1) predicts wealth holdings for whites, column (3) for blacks. Column (2) uses the white coefficient estimates with the black sample to calculate counter factual wealth holdings for blacks, column (4) the black coefficient estimates with the white sample to calculate counter factual wealth holdings for whites. Column (5) is the difference between columns (1) and (3). Column (6) is the difference between columns (1) and (2), column (7) the difference between columns (4) and (3). The percentage gap explained is in parentheses in column (6) and in column (7). It is 100 times column (6) (column (7)) divided by column (5). Standard errors account for arbitrary forms of heteroscedascity and correlation across time and persons from the same 1968 household.
Table 3
Mean regression decompositions of the race gap in wealth (siblings sample)

<table>
<thead>
<tr>
<th>Estimator</th>
<th>White coefficients</th>
<th>Black coefficients</th>
<th>Explained gap using white coefficients</th>
<th>Explained gap using black coefficients</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>White characteristics (standard error)</td>
<td>Black characteristics (standard error)</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>FE</td>
<td>64321 (1160)</td>
<td>15119 (4697)</td>
<td>17081 (603)</td>
<td>40258 (2686)</td>
</tr>
<tr>
<td>(104%)</td>
<td>(49%)</td>
<td></td>
<td></td>
<td>(56%)</td>
</tr>
<tr>
<td>FE on means</td>
<td>64281 (1446)</td>
<td>18591 (7299)</td>
<td>16963 (877)</td>
<td>27858 (3546)</td>
</tr>
<tr>
<td>(97%)</td>
<td>(23%)</td>
<td></td>
<td></td>
<td>(97%)</td>
</tr>
<tr>
<td>OLS</td>
<td>64321 (1789)</td>
<td>11666 (3733)</td>
<td>17081 (794)</td>
<td>42766 (3774)</td>
</tr>
<tr>
<td>(111%)</td>
<td>(54%)</td>
<td></td>
<td></td>
<td>(111%)</td>
</tr>
<tr>
<td>OLS on means</td>
<td>64463 (1865)</td>
<td>13350 (4874)</td>
<td>16898 (781)</td>
<td>34052 (3435)</td>
</tr>
<tr>
<td>(107%)</td>
<td>(36%)</td>
<td></td>
<td></td>
<td>(107%)</td>
</tr>
</tbody>
</table>

A. Wealth measure: Level of wealth

| FE              | 1.30 (0.02)        | 0.55 (0.09)        | 0.49 (0.02)             | 0.73 (0.07)        | 0.82 (0.03)        | 0.75 (0.11)        | 0.25 (0.11)        |
| (92%)           | (31%)              |                    |                         | (92%)              | (31%)              | (92%)              | (31%)              |
| FE on means     | 1.30 (0.03)        | 0.58 (0.15)        | 0.48 (0.02)             | 0.56 (0.10)        | 0.82 (0.03)        | 0.72 (0.19)        | 0.08 (0.12)        |
| (88%)           | (10%)              |                    |                         | (88%)              | (10%)              | (88%)              | (10%)              |
| OLS             | 1.30 (0.04)        | 0.46 (0.08)        | 0.49 (0.02)             | 0.86 (0.07)        | 0.82 (0.04)        | 0.85 (0.08)        | 0.37 (0.07)        |
| (104%)          | (45%)              |                    |                         | (104%)             | (45%)              | (104%)             | (45%)              |
| OLS on means    | 1.30 (0.04)        | 0.47 (0.10)        | 0.48 (0.02)             | 0.72 (0.07)        | 0.82 (0.04)        | 0.82 (0.10)        | 0.24 (0.07)        |
| (101%)          | (30%)              |                    |                         | (101%)             | (30%)              | (101%)             | (30%)              |

B. Wealth measure: Ratio of wealth to permanent income

| FE              | 9.49 (0.03)        | 7.60 (0.10)        | 7.13 (0.03)             | 8.86 (0.12)        | 2.37 (0.04)        | 1.90 (0.13)        | 1.73 (0.16)        |
| (80%)           | (73%)              |                    |                         | (80%)              | (73%)              | (80%)              | (73%)              |
| FE on means     | 9.45 (0.03)        | 7.70 (0.17)        | 7.12 (0.04)             | 8.88 (0.15)        | 2.33 (0.05)        | 1.75 (0.24)        | 1.76 (0.21)        |
| (75%)           | (75%)              |                    |                         | (75%)              | (75%)              | (75%)              | (75%)              |
| OLS             | 9.49 (0.04)        | 7.45 (0.10)        | 7.13 (0.05)             | 9.07 (0.09)        | 2.37 (0.06)        | 2.04 (0.08)        | 1.95 (0.08)        |
| (86%)           | (82%)              |                    |                         | (86%)              | (82%)              | (86%)              | (82%)              |
| OLS on means    | 9.46 (0.04)        | 7.42 (0.10)        | 7.11 (0.05)             | 9.07 (0.10)        | 2.34 (0.06)        | 2.03 (0.09)        | 1.95 (0.09)        |
| (87%)           | (83%)              |                    |                         | (87%)              | (83%)              | (87%)              | (83%)              |

C. Wealth measure: Log of wealth

Notes: Computed from the trimmed pooled siblings samples without sample weights (see text for details on the trimming). The regression coefficient estimates are estimated from the trimmed pooled siblings samples without sample weights. "OLS" denotes an OLS estimate, "FE" an OLS-fixed effects estimate, and "FE on means" a OLS-fixed effects estimate based on the mean of all observations for a given individual. The regression decompositions use all observations for a given individual regardless of the estimation method. See Table 3 for the definition of columns (1) through (7). Standard errors account for arbitrary forms of heteroscedasticity and family-specific serial correlation across siblings and time.
Table 4
Mean regression decompositions of the race gap in growth of wealth (5 year changes, couples sample)

<table>
<thead>
<tr>
<th>White coefficients</th>
<th>Black coefficients</th>
<th>Explained gap using white coefficients</th>
<th>Explained gap using black coefficients</th>
</tr>
</thead>
<tbody>
<tr>
<td>White characteris</td>
<td>Black characteris</td>
<td>(standard error)</td>
<td>(percent)</td>
</tr>
<tr>
<td>cs (standard error) cs (standard error)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>Total gap</td>
<td>Explained gap</td>
<td>Explained gap using white coefficients</td>
<td>Explained gap using black coefficients</td>
</tr>
<tr>
<td>(1)</td>
<td>(2)</td>
<td>(3)</td>
<td>(4)</td>
</tr>
<tr>
<td>36160</td>
<td>15366</td>
<td>8237</td>
<td>21967</td>
</tr>
<tr>
<td>(2306)</td>
<td>(5511)</td>
<td>(1685)</td>
<td>(4668)</td>
</tr>
<tr>
<td>(74%)</td>
<td>(49%)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>A. Wealth measure: Level of wealth (first difference)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1.28</td>
<td>0.58</td>
<td>0.45</td>
<td>0.70</td>
</tr>
<tr>
<td>(0.05)</td>
<td>(0.12)</td>
<td>(0.06)</td>
<td>(0.17)</td>
</tr>
<tr>
<td>(84%)</td>
<td>(30%)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>B. Wealth measure: Ratio of wealth to permanent income (first difference)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0.39</td>
<td>0.37</td>
<td>0.15</td>
<td>0.33</td>
</tr>
<tr>
<td>(0.02)</td>
<td>(0.06)</td>
<td>(0.06)</td>
<td>(0.14)</td>
</tr>
<tr>
<td>(5%)</td>
<td>(77%)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>C. Wealth measure: Log of wealth (first difference)</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
| Notes: Computed from the trimmed pooled samples without sample weights (see text for details on the trimming). The regression coefficient estimates are estimated from the trimmed pooled samples without sample weights. The dependent variable is the first difference of the indicated wealth measure between survey years. See Table 3 for the definition of columns (1) through (7). Standard errors are based on OLS formula.