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# SOCIAL CAPITAL AND HOUSEHOLD INCOME DISTRIBUTIONS IN THE UNITED STATES: 1980, 1990

by

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*October 1997*

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# SOCIAL CAPITAL AND HOUSEHOLD INCOME DISTRIBUTIONS IN THE UNITED STATES: 1980, 1990

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## Executive Summary

This report asked if changes in income inequality and the level of income were related to changes in social capital or the strength of relationships. Social capital is defined in this report as one's sympathy (antipathy) for others, idealized self, and things. Changes in social capital are expected to produce the following economic consequences. First, increases in social capital are expected to alter the terms of trade and to increase the likelihood of trades between friends and family. Second, increases in social capital are expected to increase an economic agent's concerns for the external consequences of his or her choices, internalizing what otherwise would be considered externalities. Third, increases in social capital between firms are expected to increase the likelihood that they will act in their collective interest. Fourth, increases in social capital are expected to increase the opportunities for specialization and the likelihood of trade. Finally, increases in social capital are expected to raise the average level of income and reduce the disparity of income.

This report tested empirically the relationship between changes in social capital indicator variables and changes in the average and coefficient of variation (CVs) of household income. State CVs and averages of household income were calculated for all 50 states and for different races/ethnic groups using the U.S. Census data for 1980 and 1990. Social capital indicator variables selected to measure changes in social capital included measures of family integrity including the percentages of households headed by a single female with children; educational achievement variables including high school graduation rates; crime rate variables including litigation rates; and labor force participation rates. The social capital indicator variables appeared to be significantly correlated with each other. However, in 1980, the percentages of households headed by a single female with children was not significantly related to the birth rates of single teens. By 1990, however, a strong correlation was found between the percentages of households headed by a single female with children and the birth rate of single teens.

Income inequality among U.S. households measured using CVs increased between 1980 and 1990 in all 50 states. The largest increase in CVs was among white households. The smallest increase in CVs was among Asian households. The states with the largest increase in the ratio of 1990 and 1980 CVs were Arizona, Wyoming, Maine, Vermont, and Texas. Half of the states reported decreases in real household income between 1980 and 1990. Those states with the largest percentage decrease in real income were Wyoming, Alaska, Montana, Louisiana, and West Virginia. The largest percentage increase in real income was reported by Connecticut, New Jersey, Rhode Island, and Massachusetts.

State CVs and averages of household income were regressed on four factors or subsets of social capital indicator variables. The four factors used to predict CVs and averages of household income were generally statistically significant. Thus, the findings of this report support the conclusion that changes in social capital have a significant effect on the disparity and level of household income.



# Social Capital and Household Income Distributions in the United States: 1980, 1990

## I. Introduction

Income inequality has been on the rise in the U.S. since the 1970's. Evidence of income inequality is the percentage of total income earned by the highest income families and the percentage of total income earned by the lowest income families. According to recent Census Bureau data, the 25% highest income families now receive 44.6% of U.S. income. The 25% lowest income families earn 4.4%. This is the widest rich-poor gap since the Bureau revised their data collection methods in 1947 (Bernstein).

In addition to the increase in income disparity there have been other changes in American society. Putnam (1995) claims that Americans volunteer less, are less engaged politically, have declining education standards, face rising crime rates, and have lost the sense of security due to changes in the work place. Putnam suggests that the social changes just described indicate a decline in social capital or in the quality of our relationships. Other scientists dispute Putnam's findings claiming that instead of a decay in civic engagement, we are experiencing shifting civic and community engagements (Clark).

At issue is the following question: *Do changes in social capital influence the level and disparity of household income in the United States?* The evidence presented in this paper suggests that changes in social capital and changes in income distributions are related.

### Relationships and Terms of Trade

Income and wealth levels are largely dependent on the terms of trade at which one exchanges his or her goods and services. Terms of trade can be defined as the agreement between economic agents that determines the quantity, quality, risk, price, information content, timing, and location of goods and services traded. In many experiments, relationships appear to have

altered the terms of trade. The evidence suggests that friends and family trade more and at different prices than do the estranged and strangers.<sup>1</sup> *Since relationships appear to influence the terms of trade and the terms of trade determine income distributions, then relationships must also influence income and wealth distributions.* Examples of relationships altering the terms of trade follow.

When farmland sales are recorded, a distinction is made between land sales between family members and "arms-length" sales made between unrelated individuals. The distinction is made because realtors recognize that the sale price of land depends on the relationship between the seller and the buyer (Gilliland). Nepotism laws restrict government employers from hiring their close relatives. These laws recognize the tendency of some government employers to grant employment advantages to their relatives. Civil rights laws preclude employment being denied because of one's race/ethnicity. These laws recognize that race/ethnicity often changes the relationship between employers and potential employees. Finally, our judicial system emphasizes the role of relationships by placing a blindfold on our symbol of the court, Lady Justice. The blindfold helps her make impartial judgments free from the bias created by knowing who is to be judged.

Families represent an organization in which a special relationship exists. One way this special relationship manifests itself economically is in the formation of business agreements. Gwilliam found that 89% of Michigan farmland leases were between friends and family members. Nelton noted that family businesses account for 76% of Oregon's small companies. Calonius wrote that 75% of U.S. companies are family-owned or controlled.

Relationships between individuals and causes represented by particular organizations account for substantial amounts of voluntary donations.



Despite a sluggish economy, philanthropic giving across the nation in 1991 exceeded donations in 1990 by 6.2%. Voluntary donations in 1991 equaled \$124.7 billion, of which individuals contributed 89%. The largest recipients of philanthropic giving included religious and educational organizations. Other recipients included environmental groups, the arts, health organizations, and other nonprofit groups providing human services (Tetsch).

Other studies demonstrating how relationships change the terms of trade include the following. Graduate students in the Department of Agricultural Economics at Michigan State University would sell a used car valued at \$3,000 for \$420 less than its market value if the buyer were a friend. However, these same graduate students would require \$697 above the market price if the buyer were an unpleasant neighbor (Robison and Schmid, 1991). A survey of 103 Michigan bankers serving communities of less than 10,000 found that good business and social relationships increased the probability of loan approval in some cases by 60% (Siles, Hanson, and Robison). Survey respondents reported that their willingness to bear risk depended on the consequences of their risk decision on important others (Robison and Hanson, 1996). Finally, relationships have been significant factors in customer retention, tipping behavior, data perception, and willingness to cooperate (Robison and Hanson, 1995).

### **Changes in Income Levels and Income Disparity**

It is generally agreed that income inequality may have serious social and economic consequences. Addressing the question, "What are the five biggest challenges Clinton faces in his second term in office?," Professor George J. Borjas of the John F. Kennedy School of Government wrote that Clinton's number one challenge was to address income disparity. Professor Jeffrey Rosensweig of the Goizueta Business School at Emory University wrote that Clinton's number one challenge was to ease the gulf between rich and poor and keep the middle class from hollowing out.

It has been argued by some economists that there is a tradeoff between increasing incomes and reducing income inequality. One view attributed to Kaldor is that a high level of savings is a prerequisite of growth, and since the rich save more than the poor, growth requires income be concentrated in the hands of the rich whose savings rates are high. A second view suggesting the tradeoff between income growth and income inequality is attributed to Kuznets. His view was that as labor shifts from a low productivity sector to a high productivity sector, aggregate inequality must initially increase substantially and only later decrease. This latter view, Robinson observed, has "acquired the force of economic law." The validity of this law, however, is not universally accepted. Commenting on Kuznets' law, Fields wrote:

*Perhaps one of the greater ironies in the history of thought on economic development is that the economic law which today is most often associated with Kuznets and that has come to bear his name the idea that income inequality increases in the early stages of economic development and decreases in the later stages, thus tracing out an inverted-U curve receives remarkably little empirical support, either from the evidence presented in Kuznets' writings or in subsequent data (p. 462).*

Deininger and Squire used an expanded data set covering 30 years to test for the presence of the Kuznets' curve. They found no evidence of the Kuznets' curve in almost 90% of the cases they examined.

Other explanations for the increasing income disparity include falling wages for unskilled workers as automation spreads, low tax rates on the wealthy during the 1980's, low minimum wages, the decline of trade unions, and the rapid rise in the 1980's of the stock and bond markets, in which the wealthy are heavily invested (Bradsher). More recently, Williamson suggests migrations of unskilled workers can explain a significant portion of change in income disparity.



It appears that important work remains to be done before economists and others agree on the causes of income inequality. To improve our understanding, this study considers the possibility that relationships may be a significant factor in explaining income inequality.

What follows is a report of research efforts that examine the connection between relationships and income distributions. After the introduction, sections II and III define social capital and describe its properties, including opportunities for investments and disinvestments. Sections IV and V describe how social capital internalizes externalities and alters terms of trade. Sections VI and VII describe how social capital influences income distributions and develop several hypotheses. Sections VIII through XIII test the connections between social capital and income distributions using U.S. Census data for 1980 and 1990 and indicator variables for the same period. Finally, section XIV summarizes the report.

## II. What is Social Capital?

### Relationships and Social Capital

Suppose person  $i$  perceives a change in the well-being of person or object  $j$  and as a result experiences a change in his or her own well-being. Then  $i$  is said to have a relationship with  $j$ . Besides relationships with other persons, objects of relationships may include places, communities, schools, clubs, animals, organizations such as churches and service clubs, and legal institutions. The relationship person  $i$  has with person, or object  $j$  depends on at least two elements. The first element is awareness or "social distance." Social distance measures  $i$ 's knowledge of  $j$  that may include information about  $j$ 's behavior, consumption, wealth, values, or social bonds. As  $i$ 's knowledge of  $j$  increases,  $i$ 's social distance to  $j$  decreases. For those individuals, groups, communities, or institutions  $j$  about whom  $i$  has no knowledge,  $i$ 's social distance to them is infinite and  $i$  has no relationship with them. Consequently, if  $i$  has no knowledge of  $j$ , then changes in  $j$ 's well-being do not influence  $i$ 's well-being.

The second element that determines  $i$ 's relationship to  $j$  is the degree of sympathy or antipathy that  $i$  holds toward  $j$ . Person  $i$  may develop toward  $j$  feelings of sympathy, antipathy, or neutrality (Bogardus). We have observed that stable relationships between persons tend to be symmetric; that  $i$  cares for  $j$  about the same as  $j$  cares for  $i$ .

Suppose  $i$  has awareness of and sympathy for  $j$ . Then, any improvement in  $j$ 's well-being also benefits  $i$  vicariously. As a result,  $j$  can expect person  $i$  to extend favors, preferential terms of trade, and in other ways look out for  $j$ 's interest as long as the favors, preferential terms of trade, and other benefits extended do not impose a cost on  $i$  greater than  $i$ 's vicarious benefits earned as  $j$ 's well-being improves. The relationship of sympathy (antipathy)  $i$  has toward  $j$  is called here  $j$ 's social capital with  $i$  (denoted  $k_{ij}$ ) and is defined next.

### Social Capital: A Definition

**Definition.** *Social capital is the sympathy (antipathy) one person has toward another person, idealized self, or object. The sympathetic (antipathetic) person is said to supply social capital while the person or object of sympathy (antipathy) is said to possess social capital. The persons or objects of social capital may expect benefits (harm), advantages (disadvantages), and preferential (discriminatory) treatment from the providers of social capital. Social capital may be culturally dependent, environmentally influenced, and responsive to a wide range of stimuli including the perceived social capital claimed by others.*

The above definition of social capital has benefitted from the work of an interdisciplinary team of social scientists at Michigan State University consisting of sociologists, a psychologist, a political economist, a human ecologist, a public affairs specialist, and agricultural economists.



Objects such as clubs, service organizations, corporations, communities, families, and schools may all possess social capital granted by individuals. Then, organizations endowed with social capital and other resources by individuals may establish rules and operating procedures that make its resources available to its members or others. In other words, because of social capital, institutions may endow its members with potential benefits (harms), favors (disadvantages), and preferential (discriminatory) treatment.

Other definitions of social capital include: (1) the social obligations or "connections" which are convertible into economic capital under certain conditions (Bourdieu); (2) a resource of individuals that emerges from their social ties (Coleman, 1988); (3) the ability to create and sustain voluntary associations (Putnam, 1993); (4) trust (Fukuyama); and (5) the relationship or caring between persons and between persons and their institutions (Robison and Schmid, 1994).

Coleman discussed social capital and its application to sociology.<sup>2</sup> Hyden discussed social capital in a political science setting. Putnam suggested recently that the supply of social capital in the United States has decreased. Fukuyama associated social capital with trust and suggested that trust or social capital is at the foundation of collective action. C. Flora and J. Flora discussed the importance of social capital in maintaining society's social contract. Robison and Schmid (1991, 1994), Robison and Hanson (1995, 1996), and Schmid and Robison discussed the role of social capital in economics. Finally, Evans and Fox have written about the role of social capital in development.<sup>3</sup>

### **Economic Properties of Social Capital**

Social capital and other forms of physical capital have many features in common. Like physical capital, the potential benefits of social capital can be depreciated through neglect and the passage of time. Like physical capital, social capital can sometimes be enhanced or depreciated by providing or extracting services. For example, asking a friend for a favor or improved terms of trade in an economic transaction may reduce the

likelihood the friend will extend favors in the future. On the other hand, granting favors and extending favorable terms of trade may increase one's social capital and increase the likelihood of receiving favors in the future. Finally, like financial capital, social capital may be fungible. For example, an owner of social capital may be able to alter the terms of trade for another party using his or her social capital. For example, person A may use his social capital with person B to benefit person C even though C has no social capital with B.

Perhaps one of the most interesting aspects of social capital is that it provides a new perspective on goods and markets. Traditionally, economists have described goods as objects with properties wanted by consumers. For example, a good may be wanted because of its temperature, taste, sight, place, form, location, or ability to create physical sensation. Exchange prices and amounts of the goods exchanged were then said to depend on incomes, marginal utility for the good's properties, and the cost of supplying the good. Social capital theory suggests that the desirability of a good may be modified by the relationship between the consumer and the good's supplier.

Finally, social capital theory suggests that transactions involving exchanges of money for goods may also include investments (disinvestments) in social capital. In some markets, goods may be exchanged for social capital only. For example, one neighbor may provide another neighbor an item such as a cup of sugar and refuse money as payment. The neighbor supplying the sugar may refuse payment because the increase in social capital he or she receives by supplying the sugar is valued more than the money value of the cup of sugar. In other exchanges where the good is of significant value, a good may be exchanged for both money and social capital. Consider, for example, the exchange of a used car. The seller of the used car may offer it at a discount to a friend or family member. The car seller receives less than the market value of the car plus social capital worth more than the discount offered the buyer.



Another important economic property of social capital is its ability to reduce transaction costs. High monitoring costs, threats of litigation, price, quantity, and quality discovery costs and the costs of writing contracts that consider many contingencies may all be reduced by increases in social capital between trading partners. These transaction costs are reduced by increases in social capital because each party to the trade has his well-being linked to the well-being of his or her trading partner. Thus, we expect to find less cost in writing contracts and more successful contracts between friends and family. Supporting the assumption that social capital reduces transaction costs, Johnson et al. found that farmland leases between related individuals were often oral and more successful than written leases between unrelated lessees and lessors.

The importance of social capital's ability to reduce transaction costs has important implications. In some markets, especially in less-developed countries, transaction costs are very high. As a result, opportunities for mutually beneficial trade between strangers are limited. Thus, we expect to find more trades between family and friends than between strangers, especially in high transaction cost economies.

Another economic property of social capital is its ability to change the terms of trade. Suppose person *i* has an object for sale and expects his/her well-being to be improved by exchanging the object for the object's "arm's-length" value. Now suppose that *i* has a friend *j* who needs the object being sold. Selling to *j* would improve *i*'s well-being in several ways. First, *i* would benefit from the money received from the sale. Person *i* would also receive some satisfaction from knowing *j*'s well-being has improved as a result of the sale. Person *i* may also feel good about his/her relationship to his/her idealized self, knowing that *j*'s well-being has improved as a result of his/her efforts. Finally, *i* may benefit from *j*'s improved goodwill which increased because of the purchase. Because of social capital, *i* could sell at a price below the market price and still be better off than selling the object at the market price to a stranger. Thus, because of social capital, *i* is likely to offer more favorable

terms of trade to his/her friend or family member than he or she would offer to a stranger.

Not all economic properties of social capital are beneficial economically. Consider some possible negative consequences of social capital. One negative consequence is that social capital may lead to agreements that are not economically sound. For example, a parent may employ a son or a daughter in the family business and pay them a wage higher than would be expected in a strictly "arm's-length" economic exchange. Employees may receive benefits from their employer because of social capital rather than job performance. Such job discrimination practices are resented by employees without social capital and may lead to labor unrest. In other cases, organizations may form based on certain social capital traits that exclude others. Such exclusive organizations have been called good old boys' and good old girls' clubs. Particularly improper are social capital favors extended in public work places when those extending favors are allocating public funds on criteria other than expected public benefits.

### **III. Investments and Disinvestments in Social Capital**

Economic activities between economic agents may increase or decrease their levels of social capital. Whether or not economic activities involving agents produce investments or disinvestments in social capital depend at least in part on whether or not the economic activities involving persons *i* and *j* are competitive or synergistic.

#### **Investments in Social Capital**

Investments in social capital likely occur when individuals participate in synergistic or cooperative activities. Synergistic activities are those in which one agent's success improves the likelihood of another agent's success. Synergistic activities often build social capital because the agents engaged have an interest in each other's success. Then, because they are interested in each other's success they are more likely to



communicate, join common causes, offer favorable terms of trade, share responsibility, develop emotional and social ties, and interact in still other synergistic activities, all of which increase social capital. Examples of social capital building activities include informed free trades, sharing information, transfer of gifts more valuable to the receiver than the giver, and joining and participating in service activities of service clubs, churches, schools, and professional and civic organizations.

Agents with common backgrounds and traits are more likely to engage in synergistic activities than agents who lack common traits. Some common traits and backgrounds that have been the basis of sympathetic relationships include gender, economic status, occupation or profession, common enemies, memberships in organizations, manners of dress, marital status, age, education, location of home and work, political preferences, race/ethnicity, religious preferences, geographic origin, language, national origin, moral values, and genealogy. Some traits such as race/ethnicity, family, national origin, and genealogy are durable and are therefore likely to provide a more stable basis for social capital than shared economic opportunities.

Investing in social capital through personal contact is limited by one's time and means of communicating and relating with others. Organizations can increase the efficiency of their members' efforts to build social capital in at least two ways. First, organizations may improve the efficiency of their members' social capital investment efforts by increasing their members' opportunities to meet and communicate with others. Second, organizations may improve the efficiency of their members' social capital investment efforts by establishing a clearly stated set of values to which all of its members subscribe. Then, because the organization's members know they hold similar values, they all have some social capital with each other even if they are not personally acquainted.

Finally, it appears that freedom of action is an important factor in social capital formation. Worker productivity increases when individuals

are given greater freedom of decision making. When greater freedom of decision making is granted to those within an organization, those given increased responsibility and freedom often respond with increased loyalty and greater productivity.

### **Disinvestments in Social Capital**

Disinvestment in social capital between two persons likely occurs when they participate in competitive activities, sometimes called zero sum games. In competitive activities, the goal(s) of one person cannot be achieved unless the goal(s) of the other person is (are) frustrated. Persons participating in competitive activities often develop antipathy because they view each other as threats to their own success. Activities likely to have competitive goals include athletic events, elections, divorces, estate settlements, assignment of contracts, litigation, and wars. Other examples include quarantines, embargoes, strikes, contests for budget shares, promotions, employment, and market shares.

Competitive activities not only have the tendency to reduce existing levels of social capital between groups or people, but they may also create antipathy based social capital. A disadvantage of antipathy based social capital is that it produces perverse economic behavior. The perverse behavior produced by antipathy based social capital is the willingness of an economic agent to reduce the well-being of another person or object even if it means reducing his/her own well-being. Evidence of the perverse consequences of antipathy are the 20 million people who have died in armed conflicts since the end of World War II (Korten).

Developing social capital within a group by creating antipathy between groups is a common approach for building social capital in business and politics. One reason the formation of social capital through competitive means is so popular is because of the power of antipathy. For example, in the used car sale study reported earlier, positive social capital resulted in a \$420 discount while negative social capital resulted in a \$697



premium. The negative premium was 166% of the positive discount.

Finally, disinvestment in social capital occurs when force is used to compel compliance. The force used to compel compliance may include litigation, threats of violence, deception, or physical force. One remarkable quality of the antipathy based social capital that results from the use of force is its longevity. For example, the antipathy that exists between Protestants and Catholics in Northern Ireland and between Serbians and Croats originated in conflicts involving force over hundreds of years ago.

#### **IV. What are Externalities?**<sup>4</sup>

In the discussion that follows, we intend to show that social capital rich networks have an important economic advantage not available to networks that lack social capital. This important benefit enjoyed by a social capital rich network is that they internalize many economic consequences of their actions that would otherwise be treated as externalities. Internalizing externalities improves the terms of trade for those who enjoy high levels of social capital and results in individuals acting in the interest of the group.

An externality is created when one person's action alters the well-being of another person without that person's consent or agreement. An action that increases (reduces) the well-being of another person without that person's consent or agreement is said to be a positive (negative) externality.

Externalities can be viewed in two ways. First, an externality can be viewed as a by-product of a production process. For example, there are the intended meat products of a pork production process and unintended products of the process including odor and animal waste production. Second, an externality can be viewed as an input in the production process. For example, the air that carries the odors and the water and other resources required to dispose of animal wastes are inputs into the pork production process.

If externalities are viewed as inputs in the production process, a view adopted in this report, the critical issue is who owns these inputs? Resource owners can use their inputs in the production process as they want, even if the resources are scarce and others want them for alternative uses. The pork producer uses the air to carry away odors. The residents desire the air without the odor for breathing but because they cannot claim exclusive ownership, they cannot restrict the pork producer's use of air in pork production. Residents near the pork producer may bid away the use of the resource either through purchase or by establishing legal claims. Yet often, residents desiring to own the clean air may be many and no one or few alone can offer sufficient bids to the pork producer to induce him or her to alter the use of the clean air.

The production of externalities is related to social capital in at least two ways. First, as those who lack ownership of resources increase their social capital provided by resource owners, production plans and terms of trade are altered. Because of the increased social capital, owners of resources that may be used in ways that produce positive (negative) externalities may defer their use of the limited resource to others. Alternatively, caring owners may use their resources less intensively and by so doing reduce the externalities for others.

The second consequence of social capital on the creation of externalities is related to property rights. Ownership rests on a consensus of legitimacy. The willing acceptance of another's right to the opportunities of ownership requires some minimal threshold of respect, if not care, for the owner. A despised owner is an insecure owner (Schmid). Expenditures to secure ownership against those who contest or deny ownership range from payments to police, guerrilla groups, lawyers, and armies. As social capital increases, expenditures required to secure ownership rights decrease and can be used more productively. It may be the case that the most serious impediment to economic development is the high cost of enforcing property rights that divert resources from the production of goods and services.



Consider some examples of economic actions that create externalities. Residents pay taxes to support public education even when they have no children in the school system. While there are some direct benefits to childless voters from living in a community with better educated children, many would describe the basis of their school support as their interest in the well-being of the community's children. Citizens vote for bonds that provide for themselves fire and police protection. However, not all community members pay the same for the nearly identical protection they all enjoy. In these examples, citizens are extending resource ownership rights to other citizens who may contribute differently to the creation of the resource.

Citizens who obey the law without compulsion reduce law enforcement costs which lower taxes for them and the entire community. Most large cities have mass transportation systems that are only partially funded by those who use the system. The remaining funds supporting the transportation system are provided by nonuser taxpayers. Public radio provides services to listeners regardless of their donation to the radio station. Those who contribute to public radio create positive externalities for those who also listen but do not contribute funds for its support.

Businesses may agree to construct a mall knowing that locating next to each other lowers transaction costs for all those who shop at the mall. A customer may enter the mall for a particular purchase but while in the mall shop at other stores because of their convenient location. So, in a way, businesses create externalities for each other by bringing customers to the mall. Besides lowering transaction costs for customers by placing businesses together in a mall, the businesses often reduce the overhead costs and some variable costs for each other. Cleaning, parking, protecting, and advertising costs are shared by the businesses in the mall and are less than if each store on its own acquired the same services.

Examples of negative externalities abound as do positive ones. Often, negative externalities

involve the production of a profitable product for the producer that diverts an input from an alternative use wanted by a different group. The producer and the consumer may both benefit from the product sold and consumed. However, others not involved in the production or consumption of the desired product are often adversely affected by having to forego their preferred use of the input. Resources often sacrificed in the production of negative externalities are clean air to the production of smoke, serenity to the production of noise, passable roads to traffic congestions, clean water to polluted water that carries away wastes, and healthy soils to soils polluted with toxic wastes.

Crime is an unlawful interference with the rights of resource ownership. If the rights of ownership are considered an input into the production process, denying an owner the use of his/her resources by means of a criminal act is the creation of an externality even though the criminal may realize an economic benefit from his crime at the expense of the victim who loses his property rights. In addition, the entire community suffers an externality as a result of the crime. The community must now spend more on crime prevention and pass more restrictive laws that limit the activities of its citizens. Finally, crimes impose negative externalities on the community in the form of a lost sense of trust and security.

Clearly, the list of positive and negative externalities could be expanded. Indeed, it may be more difficult to identify activities without external effects than to identify activities with only internal effects. The destruction of rain forests in Brazil, monetary policies carried out in Mexico, and ethnic conflicts in Rwanda may once have had little effect on the U.S. economy, but no longer, as the world becomes increasingly economically interdependent.

Recognition of externalities and their importance leads us to the question: how does social capital affect the production of externalities? We will now explore the connection between social capital, the creation of negative and positive externalities, and household income distributions.



## V. Social Capital, Externalities, and Income Distributions

Having considered how social capital internalizes externalities and changes the terms of trade, we now explore the connections between social capital, externalities, and income distributions.

Consider an economy consisting of two economic agents  $i$  and  $j$ . Assume that agent  $i$  earns more income than agent  $j$ . Also assume in this hypothetical economy that agent  $i$  is engaged in economic activities that produce externalities. Then, as  $j$ 's social capital increases,  $i$  internalizes his/her externalities with the following effect on the distribution of income (see Appendix A).

*If agent  $j$ 's social capital available from agent  $i$  increases, then the combined incomes of agents  $i$  and  $j$  will increase and the difference in their incomes will decrease.*

### Four Externality Models

The importance of the connection between changes in social capital and changes in the level and disparity of incomes can be shown using four different externality models. The four externality models include: (1) the high exclusion cost good model; (2) the joint production model; (3) the goods owned in common model; and (4) the ubiquitous externality model.

*High exclusion cost good model.* A high exclusion cost good is one that allows agents to extract services from the good independent of agents' contributions to the creation of the good. A high exclusion cost good exists because of the cost of "fences" or the cost of denying access to the good to those who have not paid for its production. Examples of high exclusion cost goods include: street lights, radio programs, dams providing downstream flood protection, extensive parks with many points of entry, water sanitation plants, and neighborhood police protection. The income distribution conclusion described earlier implies that investments in the high exclusion cost

good by agent  $i$  will increase as  $j$ 's social capital increases. Because of  $i$ 's increased investment in the high exclusion cost good, the combined incomes of agents  $i$  and  $j$  will increase and the difference in their incomes will decrease.<sup>5</sup>

*Joint production model.* A joint production model is one in which production depends on inputs supplied by more than one economic agent.

Economic agents often engage in the production of both individual and jointly produced goods. For example, many vegetable farmers produce their crops individually but join with others to transport, store, and market their produce. The income distribution conclusion implies that production of jointly produced goods will increase as  $j$ 's social capital with  $i$  increases. Because of increased production of jointly produced goods, the combined incomes of agents  $i$  and  $j$  will increase and the difference in their incomes will decrease.<sup>6</sup>

*Goods owned in common model.* A good owned in common is one for which several agents have service extraction rights. The marginal cost of service extraction from goods owned in common depends on the total services extracted. Examples of goods owned in common include wildlife populations, public lands used for grazing, fishing waters, public parks, and publicly owned roads. The income distribution conclusion implies that agent  $i$ 's service extraction and exploitation of the good owned in common will decrease as  $j$ 's social capital with  $i$  increases. Because of  $i$ 's decreased use of the good owned in common, the combined incomes of agents  $i$  and  $j$  will increase and the difference in their incomes will decrease.<sup>7</sup>

*The ubiquitous externality model.* Production often involves the use of inputs that have incompatible uses. Agent  $i$  may use inputs to increase his profit but in the process preclude their use by agent  $j$ . For example, inputs necessary for pork production include land, buildings, feed, and a place to put waste. However, the water, air, and land used to handle  $i$ 's waste may be desired by  $i$ 's neighbors for other purposes. The income distribution conclusion implies that agent  $i$ 's use



of resources with incompatible uses will decrease as  $j$ 's social capital with  $i$  increases. Because of  $i$ 's decreased use of resources with incompatible uses, the combined incomes of agents  $i$  and  $j$  will increase and the difference in their incomes will decrease.<sup>8</sup>

### Externalities and Income Transfers

So far the linkages between social capital, externalities, and the income distribution of agents  $i$  and  $j$  have been described in production models. In many business arrangements, this linkage between voluntary economic actions and income distribution possibilities may be accurate. However, in most advanced economies, there exist income redistribution possibilities besides voluntarily altering production arrangements. One means of redistribution is an income transfer. Whether the transfer is voluntary or involuntary influences in different ways an agent's production decisions and the resulting externalities.

Appendix B deduces an important conclusion regarding income transfers, social capital, and difference in incomes. The conclusion deduced is that:

*If agent  $i$  because of his/her superior income position relative to agent  $j$  is forced to transfer income to agent  $j$ , then agent  $i$  will reduce (increase) his/her production of positive (negative) externalities.*

The implication is that externally imposed income transfers will be offset to some degree by production decisions with external consequences.<sup>9</sup>

Another conclusion deduced in Appendix B is that:

*If the income transfers are voluntary and agent  $i$  chooses an amount of his/her income to transfer to agent  $j$  that maximizes his/her own utility, then transfers to agent  $j$  will increase with increases in agent  $j$ 's social capital.*

These conclusions about income transfers and social capital have some important implications including the following. Externally imposed transfers intended to reduce income disparities may have their effects canceled by agents' voluntary production and investment responses. These offsetting income distribution effects should serve as warnings to social planners who believe income inequities can be eliminated with involuntary transfers. On the other hand, awareness of social capital and its usefulness in reducing income disparities may provide policy makers an important new approach for reducing income disparities. The new approach is to design programs to increase social capital.

## VI. Social Capital, Specialization and Trade, and Income Distributions

To learn more about the relationship between changes in social capital and changes in the level and disparity of income, consider income distributions for  $N$  individuals or firms instead of two firms or individuals  $i$  and  $j$  used in the earlier deductions. In an earlier section, it was pointed out that trades are likely to increase with increases in social capital. Trading, of course, has the desirable economic outcome of permitting economic agents to specialize in a particular economic activity. One fundamental tenet of economics is that labor specialization increases productivity. The famous example of this point is Adam Smith's observation that a pin maker working alone could barely produce a pin a day. But, 10 pin makers working together and specializing in different parts of the pin production process could together produce up to 48,000 pins daily. The advantage of labor specialization is that one's ability to perform a task is often improved through repetition. Second, labor specialization allows one to participate in economic activities for which the agent is best suited. But labor specialization cannot occur without trading since a specialization means agents must give up the production of desired goods which must then be acquired through trading. Thus, trading and labor specialization are linked in any economic system.



Increases in social capital increase the incentives for labor to specialize and trade by internalizing the benefits of trade received by one's trading partner. Moreover, since social capital is most likely to develop between family and friends, among these are most likely to develop trading relationships. As improvements in social capital increase the size of the trading group, additional opportunities for labor specialization and trade are created. Then, with increases in trades and labor specialization, the average productivity of group members increases (one pin versus 48,000 pins). The final result of increased social capital and increased group size is that the average level of income increases.

### Trading Opportunities and the Distribution of Income

Recognizing that social capital opportunities influence trading opportunities, we next consider how changes in social capital may change the distribution of income. To begin, suppose the world is organized by countries (firms or households) and that each country produces one unique product for export (trade). Next, suppose that barrier-free trade exists among all  $N$  countries allowing  $T_0 = N(N-1)/2$  pairs of trading arrangements to develop. As a result, each country would benefit from the productive skill of the other  $N-1$  countries and enjoy the opportunity to consume some of their exports. In addition, they might combine imported products to create new products. More complicated goods require more imported products to produce. Less complicated goods require fewer imported products to produce. Finally, we might also assume that under conditions of perfect social capital, total income would be evenly distributed among the  $N$  countries.

Next, suppose that the world of countries is divided into two groups of equal size. Assume also that near-perfect social capital exists within the two groups of countries but that antipathy exists between the two groups. Because of antipathy, it is assumed that trading is impeded between countries in the two different groups. In a world divided into two groups (one division), the number of unique trades available to each

country is reduced to  $T_1$ . Furthermore, the rates of  $T_1$  to  $T_0$ , is equal to:

$$\lim_{N \rightarrow \infty} \frac{T_1}{T_0} = \frac{\frac{1}{2} - \frac{1}{N}}{1 - \frac{1}{N}} = \frac{1}{2} \quad (1)$$

The implication of equation (1) follows. Dividing  $N$  countries into two groups of equal size reduces each country's potential trading partners from  $(N-1)$  to  $(N/2-1)$  and reduces by 50% the total number of different goods that could be produced using two unique inputs. In addition, using an equation similar to equation (1) it is easy to demonstrate that one division of  $N$  countries into two equal groups reduces by 75% the total number of different goods that could be produced using three unique inputs and reduces by 87% the total number of different goods that could be produced using four unique inputs. Other ratios of unique goods produced before and after divisions depending on the number of inputs required are described in Table 1.

It may be the case that not all of the  $N$  countries in our model produce unique products. If this were the case, the reduction in the total number of different goods produced as a result of dividing the  $N$  countries into groups of equal size would be less dramatic than the results described in Table 1. Still, there is an important lesson to be learned. It is that a loss in social capital that leads to divisions and trade barriers between previously unified groups decreases dramatically the production of processed or complicated goods. As social capital increases and the number of groups decrease, then the number of trading partners and opportunities to specialize also increases. The result of increased specialization and trade, as Adam Smith demonstrated, is to increase the productivity of labor.

Finally, as the productivity of labor increases with increases in the membership of the social capital rich group, the mean income of the group can be expected to rise. In addition, as membership in the social capital rich group



increases, externalities are internalized for an increasing number of economic agents and the results deduced in Appendix A apply. These results suggest that for the externality models already discussed, mean income will increase with an increase in the membership of the group and the disparity of income will decrease.

Whether or not the mean income increases linearly or increases at a decreasing (increasing) rate as the size of the social capital rich group increases is an empirical question. Opportunities for trade within the group increase at an increasing rate as the size of the group increases. So we make the assumption that income increases linearly or at an increasing rate with increases in  $N$ . On the other hand, as the size of the group increases, the demand for bonding activities may also increase at an increasing rate. In addition, the cost of maintaining social capital as the group size increases may effectively limit the size of the group unless efficient means of investing in social capital are introduced. One means for efficient social capital investments already mentioned was to establish organization based on commonly accepted values.

The discussion about group size has

emphasized the advantages of trade. What makes the discussion relevant is an important empirical fact. In 1945 when the United Nations was founded, the world was organized into 51 countries. This number increased to 100 in 1960. By the year 1994, the number of countries had increased to 192. Since 1994, the number of countries has continued to increase (Bradshaw and Wallace). If increasing the number of countries results in trade restrictions between those who were formerly members of the same country, then we can expect the consequences just described; mainly, less labor specialization and reduced income for each group member.

### Social Capital, Specialization and Trade, and Kuznets' Law

Suppose that there exists an economy of  $N$  households, all perfect and symmetrically endowed with social capital so that each values each other's income the same as his or her own. Furthermore, assume that each household earns  $y(N)$  income which increases with increases in  $N$ . This arrangement represents the ideal and should result in the highest level of equally distributed income.

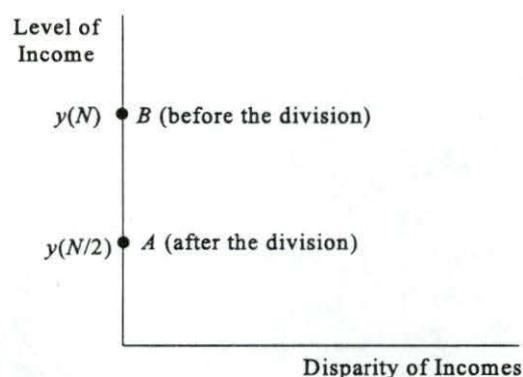
**Table 1. Percentage of Goods Produced After Division(s) Compared to the Original Number of Goods Produced Before Division(s)**

| Number of Groups After Division(s) | Number of Unique Inputs Required Per Good Produced |     |     |     |     |     |     |     |      |
|------------------------------------|--|-----|-----|-----|-----|-----|-----|-----|------|
|                                    | (2)  | (3) | (4) | (5) | (6) | (7) | (8) | (9) | (10) |
|                                    | Percentages  |     |     |     |     |     |     |     |      |
| 1                                  | 100  | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100  |
| 2                                  | 50   | 25  | 13  | 6   | 3   | 2   | 1   | 0   | 0    |
| 3                                  | 33   | 11  | 4   | 1   | 0   | 0   | 0   | 0   | 0    |
| 4                                  | 25   | 6   | 2   | 0   | 0   | 0   | 0   | 0   | 0    |
| 5                                  | 20   | 4   | 1   | 0   | 0   | 0   | 0   | 0   | 0    |
| 6                                  | 17   | 3   | 0   | 0   | 0   | 0   | 0   | 0   | 0    |
| 7                                  | 14   | 2   | 0   | 0   | 0   | 0   | 0   | 0   | 0    |
| 8                                  | 13   | 2   | 0   | 0   | 0   | 0   | 0   | 0   | 0    |



Next, suppose that a dispute arises that divides the  $N$  households into two groups. Assume also that the division destroys the social capital between the two groups but within the two groups, social capital remains perfect and symmetrically distributed. Under this new arrangement, one might argue that incomes remain equal within groups since perfect and symmetrical social capital exists. Furthermore, if the average income levels of the two groups are equal, then income must be evenly distributed for all  $N$  households just as it was before the division. What is different from the first case is that the opportunities for specialization and trade have been reduced. In addition, within the two groups externalities are internalized to a lesser degree than when there was only one group. The result is that the average level of income has been reduced from  $y(N)$  to  $y(N/2)$ .

Suppose that the means and the disparity of income measures associated with the household income distributions were plotted. The two distributions represented by their means and disparity measures would be represented as two points on the vertical scale that measures the means of household income for zero variations in household income. The point representing the income distribution before the division is described as point B in Figure 1. The point representing the income distribution after the division is described as point A in Figure 1. The two distributions are distinguished only by differences in their means, providing one example of how a decrease in social capital changes the mean but may leave the disparity of income unchanged.

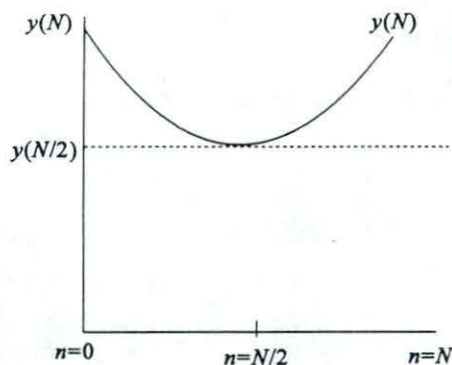


**Figure 1. The Distribution of Incomes Before and After a Division**

Next, consider the consequences on the distribution of incomes between the two groups if membership in the two groups were not evenly divided and the size of the group was the sole determinant of the level of income. Consider first changes in the overall mean level of incomes earned in the two groups as membership in the first group,  $n$ , grows from zero to  $N$  while membership in the second group,  $(N-n)$ , decreases from  $N$  to zero. With  $n=0$ , all  $N$  members of the population earn the highest level of income possible because the size of the second group is at its maximum. Nevertheless, as  $n$  increases to  $n=N/2$ , the overall mean of income earned by members of the two groups decreases to  $y(N/2)$ .

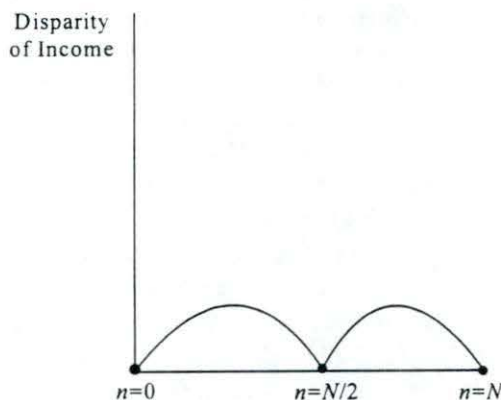
The reason mean income decreases as  $n$  grows to  $N/2$  is because members of the larger group are joining the smaller group. As a result, they are exchanging a higher income for a lower income. Moreover, all members of the larger group suffer a loss in income while all members of the smaller group earn a higher income. Nevertheless, for  $n < N/2$  the number of persons suffering a reduction in income is greater than the number of persons enjoying an increase in income. As  $n$  increases past  $N/2$  in size, the mean income increases to its original value obtained when  $n=N$ . Thus, the mean income produces a "U" shape pattern as  $n$  increases. The "U" pattern of mean income in response to changes in  $n$  is described in Figure 2.





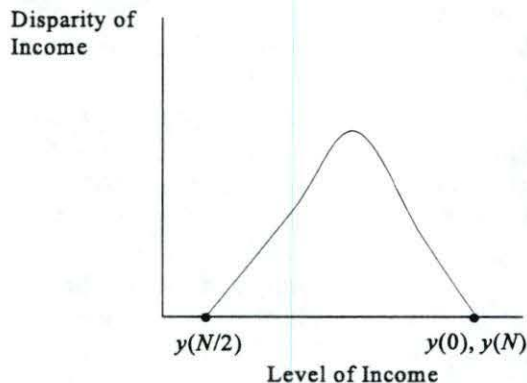
**Figure 2. The Effect on Average Income of Changing Group Sizes**

The effect of an increase in  $n$  on the disparity of income is more complicated than the effect of an increase in  $n$  on mean incomes. Maintaining our assumption that the mean income depends on the size of one's group, then the disparity of income is zero for  $n=0$ ,  $n=N$ , and  $n=N/2$ . In the first two cases, all  $N$  members of the population belong to one or the other of the two groups. For the third case, all  $N$  members of the population belong to groups of equal size and therefore earn equal incomes. Thus, the disparity of income first increases and then decreases toward zero as  $n$  approaches  $N/2$ . Then, as  $n$  increases in size beyond  $N/2$ , the pattern is repeated. The relationship between income disparity and increases in  $n$  is described in Figure 3.



**Figure 3. The Effect of Group Size on Disparity of Income**

Income distributions measures described in terms of their means and disparity of income for increases in  $n$  are described in Figure 4 by combining Figures 2 and 3. Note that the disparity of income first increases and then decreases with increasing means of income, reminiscent of the earlier described Kuznets' law.



**Figure 4. The Result of Changes in Group Size on Average Income and Disparity of Income**

## VII. Distribution of Income and Social Capital

The results of the previous section can be generalized by configuring social capital groups in another way while retaining the restriction that the overall population is  $N$ . For example, assume there exists groups of different sizes and that incomes in each group depend on the number of members. (We continue to assume that income within groups is equally divided.) To describe a particular distribution of income, let  $y(n)$  represent the income earned by each member of a group of size  $n$ . Moreover, let  $f(n)$  represent the number of groups of size  $n$  that exist in the current distribution. Of course, the restriction must be imposed that  $N$  equals the sum over all possible  $n$  values of  $nf(n)$  or  $N = \sum_n nf(n)$ . The population

distribution according to its respective group size is described by the function:

$$g(n) = \frac{nf(n)}{N} \quad (2)$$



which if summed over all possible values of  $n$  equals one.

The importance of the function  $g(n)$  is that it has all the properties of a probability distribution. Moreover, by treating  $g(n)$  as a probability density function, we can calculate the mean and a disparity of income measure for each observed income distribution.

In our economies of social capital rich groups that restrict trading activities to members of their group, the mean income of the population is described by the function:

$$\mu_y = \sum_{n=1}^{\infty} y(n)g(n) \quad (3)$$

The average variation in income from the mean, an income disparity measure, can be measured by the standard deviation of the income distribution. The standard deviation equals:

$$\sigma_y = \left\{ \sum_{n=1}^{\infty} [y(n) - \mu_y]^2 g(n) \right\}^{1/2} \quad (4)$$

A measure of disparity closely related to the standard deviation is the coefficient of variation (CV) which measures the average percentage deviation from the mean. The CV measure for the distribution of household income is defined as:

$$CV_y = \frac{\sigma_y}{\mu_y} \quad (5)$$

The hypotheses deduced earlier in this paper involve two dimensions of the distribution of income: the mean and the standard deviation or CV. Both the mean and the standard deviation are subject to scaling differences in the cost of living. This scaling difference, if linear, can be removed from the standard deviation by dividing it by the mean, creating the CV. Nevertheless, dividing the

standard deviation by the mean to obtain an average percentage measure of disparity from the mean changes the unit of measure from dollars to percentages.

If the so-called Kuznets' law holds empirically, we should expect to find the standard deviations of household income first increasing and then decreasing with increases in mean income. Even if means and standard deviations of income are related in the way just described, this does not imply that means and CVs are related in the same way. Suppose that means and standard deviations of household income are positively related. Suppose also that mean incomes are increasing faster than standard deviations of income. If mean incomes are increasing faster than standard deviations of income, then an inverse relationship will exist between means and CVs even though a positive relationship exists between the means and standard deviation of income. Thus, the relationship between level and disparity of income may depend on the choice of the disparity of income measure used.

### Changes in Social Capital and Income Distribution

The question we consider next is, does there exist a correspondence between changes in social capital and income distributions? To answer this question, let  $k = k_{ij}$  represent each member's social capital provided by each of the  $(n-1)$  members of his/her group. Then each member's total stock of social capital depends on his/her group size  $n$  and can be expressed as:  $h(n) = (n-1)k$ . Since the percentage of the population enjoying a social capital stock of  $h(n)$  is  $g(n)$ , we can calculate the mean and standard deviation of social capital in the population by substituting  $h(n)$  for  $y(n)$  in equations (3) and (4). The resulting expressions for the mean,  $\mu_s$ , and standard deviation,  $\sigma_s$ , of social capital can be expressed as:

$$\mu_s = \sum_{n=1}^{\infty} h(n)g(n) \quad (6)$$

and:



$$\sigma_s = \left\{ \sum_{n=1}^{\infty} [h(n) - \mu_s]^2 g(n) \right\}^{1/2} \quad (7)$$

Earlier, we described how each member of a social capital rich group enjoys an increase in income as the size of his/her group increases. Similarly, each member of the group would enjoy an increase in social capital as  $n$  increases. If there were only two groups, one of size  $n$  and the other of size  $(N-n)$ , average income and average social capital for the entire population  $N$  would first decrease and then increase as  $n$  increased from zero to  $N$  (see Figure 2). Since average income and average social capital respond similarly to increases in  $n$ , we have some reason to expect the correlation between average social capital and average income to be highly correlated. In addition, if  $y(n)$  were linear in  $n$ ,  $\sigma_s$  and  $\sigma_y$  would differ only by a constant and  $h(n)$  and  $y(n)$  would be perfectly correlated.

### Households Headed by a Single Parent with Children and Kuznets' Law

The social unit most likely to experience near-perfect social capital and the unit most likely to internalize externalities is the family or household. Supporting evidence for this conclusion is the dominance of family businesses. However, the evidence presented in this paper is that not all households enjoy the same level of social capital.

According to the U.S. Bureau of the Census, median income for married-couple families with children less than 18 years of age was \$22,568 in 1980 and \$40,693 in 1989. In contrast, median income for households headed by a single female with own children less than 18 was \$8,002 in 1980 and \$12,485 in 1989. The evidence is that households headed by a single female with own children are economically disadvantaged compared with households headed by a married couple with own children.

If social capital available in single-parent households,  $h(s)$ , is less than that available in two-parent households,  $h(m)$ , then the trends in

Table 2 should be of some interest. In 1970, single-parent families with children represented 11% of all families with children. By 1980, 19.5% of the families with children were headed by a single-parent and by 1990 the percentage had reached 24%.

To describe the effects on the level and disparity of household income associated with increases in households headed by a single parent with children, consider the following argument. Suppose there exists an economy with households that all enjoy perfect and symmetric social capital within the household. Also assume that the households enjoy a social capital resource with persons outside the household unit that depends on whether one or two parents are present as well as the size of the household and the age of the members of the household.

Next, we set the number of households headed by a single parent equal to  $p_s$  and the number of households headed by married parents equal to  $p_m$ . Let the average income of the married household be  $y_m$  and let the average income of the single-parent household be  $y_s$  where  $y_m > y_s$ . The average household income based on the assumptions and symbols just adopted equals:

$$\mu_y = \frac{p_m y_m + p_s y_s}{p_m + p_s} \quad (8)$$

And if we substitute for  $p_s$  the expression  $(N - 2p_m)$ , we can rewrite  $\mu_y$  as:

$$\mu_y = \frac{p_m y_m + (N - 2p_m) y_s}{N - p_m} \quad (9)$$



**Table 2. Families by Presence of Own Children Under 18, 1970 to Present (number in thousands)**

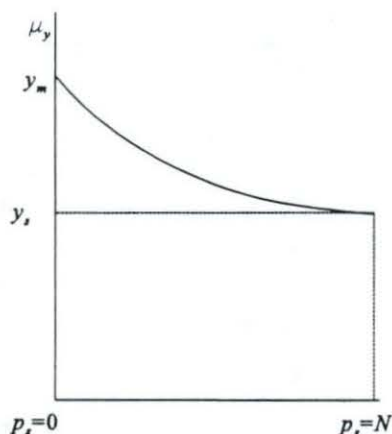
| Year              | All Families | Total Families with Children Under 18 | Families with Children Under 18                     |             |             |                         |
|-------------------|--------------|---------------------------------------|---|-------------|-------------|-------------------------|
|                   |              |                                       | One-Parent Families                                 |             |             | Married Couple Families |
|                   |              |                                       | Total Single-Parent Families with Children Under 18 | Mother Only | Father Only |                         |
| 1995              | 69,305       | 34,296                                | 9,055   | 7,615       | 1,440       | 25,241                  |
| 1994              | 68,490       | 34,018                                | 8,961   | 7,647       | 1,314       | 25,058                  |
| 1993              | 68,144       | 33,257                                | 8,550   | 7,226       | 1,324       | 24,707                  |
| 1992              | 67,173       | 32,746                                | 8,326   | 7,043       | 1,283       | 24,420                  |
| 1991              | 66,322       | 32,401                                | 8,004   | 6,823       | 1,181       | 24,397                  |
| 1990              | 66,090       | 32,289                                | 7,752   | 6,599       | 1,153       | 24,537                  |
| 1989              | 65,837       | 32,322                                | 7,587   | 6,519       | 1,068       | 24,735                  |
| 1988              | 65,133       | 31,920                                | 7,320   | 6,273       | 1,047       | 24,600                  |
| 1987              | 64,491       | 31,898                                | 7,252   | 6,297       | 955         | 24,646                  |
| 1986              | 63,558       | 31,670                                | 7,040   | 6,105       | 935         | 24,630                  |
| 1985              | 62,706       | 31,112                                | 6,902   | 6,006       | 896         | 24,210                  |
| 1984              | 61,997       | 31,046                                | 6,706   | 5,907       | 799         | 24,340                  |
| 1983              | 61,393       | 30,818                                | 6,455   | 5,718       | 737         | 24,363                  |
| 1982              | 61,019       | 31,012                                | 6,547   | 5,868       | 679         | 24,465                  |
| 1981              | 60,309       | 31,227                                | 6,300   | 5,634       | 666         | 24,927                  |
| 1980 <sup>R</sup> | 59,550       | 31,022                                | 6,061   | 5,445       | 616         | 24,961                  |
| 1980              | 58,426       | 30,517                                | 5,949   | 5,340       | 609         | 24,568                  |
| 1979              | 57,804       | 30,371                                | 5,857   | 5,288       | 569         | 24,514                  |
| 1978              | 57,215       | 30,369                                | 5,744   | 5,206       | 539         | 24,625                  |
| 1977              | 56,710       | 30,145                                | 5,270   | 4,784       | 486         | 24,875                  |
| 1976              | 56,245       | 30,177                                | 5,067   | 4,621       | 446         | 25,110                  |
| 1975              | 55,712       | 30,057                                | 4,888   | 4,404       | 484         | 25,169                  |
| 1974              | 55,053       | 29,750                                | 4,472   | 4,081       | 391         | 25,278                  |
| 1973              | 54,373       | 29,571                                | 4,184   | 3,798       | 386         | 25,387                  |
| 1972              | 53,296       | 29,445                                | 3,963   | 3,598       | 365         | 25,482                  |
| 1971              | 52,227       | 28,786                                | 3,695   | 3,365       | 331         | 25,091                  |
| 1970 <sup>R</sup> | 51,586       | 28,812                                | 3,271   | 2,971       | 345         | 25,541                  |
| 1970              | 51,237       | 28,665                                | 3,260   | 2,925       | 335         | 25,406                  |

<sup>R</sup> Revised data.

Source: U.S. Bureau of the Census.

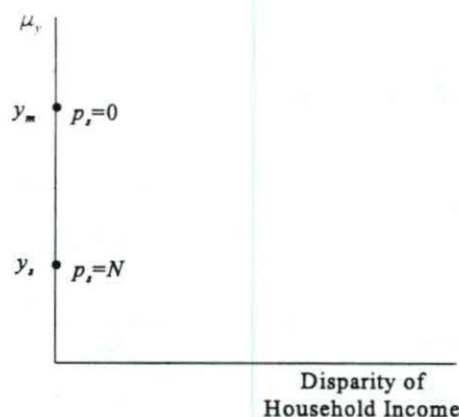


It should be apparent that if the number of households headed by a single parent with children increases, the average income of all households decreases. This result occurs because households are moving from a higher to a lower earning category. The inverse relationship between the average income and the number of single-parent households is described in Figure 5.



**Figure 5. The Inverse Relationship Between the Number of Households Headed by a Single Parent with Children and Average Household Income**

Next consider the consequences on the disparity of incomes between households as the number of households headed by a single parent increases. If  $p_s=0$ , then all households would earn  $y_m$  level of income and the disparity of income between households would be zero. Furthermore, if all households were headed by a single parent,  $p_s=N$ , then all households would earn  $y_s$  and again the disparity of income would be zero although the income level would be reduced from  $y_m$  to  $y_s$ . These two possible income distributions are described in Figure 6.

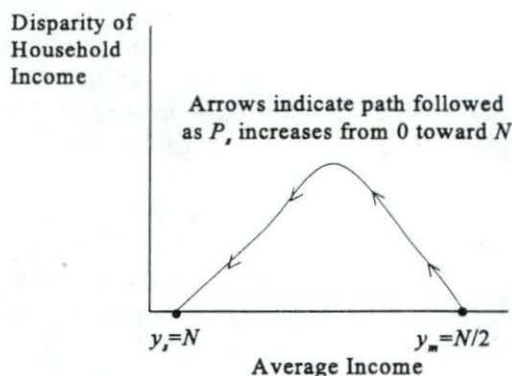


**Figure 6. The Level and Disparity of Household Income if All Households Were Headed by Either a Single or Two-Parent Family**

As  $p_s$  increases from zero to  $N$ , the disparity of income would first increase from zero and after some point would decrease until disparity of income was again zero. The relationship between increases in  $p_s$  and average household income and the disparity of household income is described in Figure 7.

The relationship described in Figure 7 is that described by Kuznets' law depicted in Figure 4. In addition, if the relationships described in Figures 4 and 7 are correct, then the level of income and disparity of income may be positively or negatively correlated. On the other hand, if the relationship in Figures 4 and 7 were limited to points where the level of income and the disparity of income were inverse, then we would observe only a negative correlation.





**Figure 7. The Effect of Increases in  $p_s$  on the Average Household Income and the Disparity of Household Incomes**

We expect that current observations corresponding to increases in the percentage of households headed by a single parent are primarily located along the portion of the graph in Figure 7 in which the level and disparity of income are inversely related. Thus, we expect to find an inverse relationship between disparity and level of household income.

### **Summarizing the Effects of Social Capital on Income Distributions**

So far, the effects of changes in social capital on income distributions have been deduced using two different approaches. The first approach used production models to show how social capital internalized externalities and increased the level and reduced the disparity of incomes. The second approach emphasized how social capital organized trade among social capital rich groups. Moreover, since group size determined opportunities for trade and labor specialization and the extent to which externalities were internalized, income per group member was assumed to increase with group size.

Finally, the report showed how the language of statistics can be used to describe income and social capital distributions. It was demonstrated that almost any pattern of income distribution can

be derived by changing the composition and number of social capital rich trading groups.

## **VIII. Indicator Variables and Household Income Distributions**

### **Indicator Variables**

Social capital indicator variables expected to be related to social capital levels include crime rates, infant mortality rates, school dropout rates, poverty rates, labor force participation rates, drug use rates, divorce rates, birth rates of single teens, rates of voluntary giving and community service, and rates of memberships in community organizations including attendance at religious services (Putnam, 1993). In the discussion that follows, some of these and other related indicator variables are organized into four groups: (1) family integrity; (2) educational achievements; (3) crime; and (4) labor force participation. In addition, we consider the effects of transfer payments on income distributions.

*Family integrity.* In her presidential address to the Rural Sociological Society, Bokemeier writes: "families and households are the critical and strategic social organization through which individuals shape and adapt to social transformations" (Bokemeier, p. 5). We assume families are the primary unit of organization in society and most responsible for the production of social capital.

The U.S. Bureau of Census (1990) defines a family as a domestic group of two or more people united by bonds of blood, adoption, or marriage. A household, in contrast to a family, is defined as a unit of co-residence (Bokemeier, p. 12). In today's North American culture, many different family and household organizations are emerging including blended families that include new spouses, ex-spouses and their parents and partners, children, and step children, cohabiting households with adults and/or children, and single and unwed parents.



In the empirical section of this paper, we are concerned with a household type that is increasing in importance, the household headed by a single female with children.

McLanahan and Booth note that those families headed by single women are at a significant disadvantage given the persistent gender gap in wages, low child support payments, and reduced access to social and cultural capital. In addition, two parents have access to the social capital associated with the extended family units of each parent that can be made available to the children. Single-parent families often lack the social capital resources of the estranged parent's family. For all these reasons, we expect the decline in families headed by two parents to reflect a reduction in social capital per household and hence to adversely affect the income distribution (see Table 2).

Whitehead supports the view that households headed by a single parent may be disadvantaged. Whitehead writes: "Children in single-parent or step parent families are more likely than children in intact families to be poor, to drop out of school, to have trouble with the law, to do worse, in short, by most definitions of well-being" (1993).

The reduced social capital resources available to children in single-parent homes do not imply that a single parent cares less for his/her children than do married parents. The assumption of reduced social capital in single-parent homes means that, in general, a single parent simply has less social capital resources to share with his/her children than does a married couple. Indicator variables selected to measure social capital associated with households include: percentage of households headed by a single female with children, birth rates of single teens, and infant mortality rates.

An important change is occurring in the cause of the formation of households headed by a single parent. A household headed by a single parent with children may arise from divorce, death, or unwed births. The important trend is the increasing importance of unwed births in the creation of households being headed by a single

parent with children. In 1980, the marital status of children living with their mothers indicated that 15% of the mothers had never married. In 1990, the percentage of mothers with children under 18 who had never married increased to 31%. In 1995, this percentage had reached 35.5%.

One fact that remains unchanged over time is that predominantly females head single-parent households with children. In 1980, 91.5% of the households headed by a single parent with children under 18 were headed by a female. In 1990, 87.4%, and in 1995, 87% of the households headed by a single parent with children under 18 were headed by a female (see Table 3).

Indicator variables selected to measure social capital associated with family integrity used in this study include: percentages of households headed by a single female with children, birth rates of single teens, and infant mortality rates.

*Educational achievements.* Some studies support the conclusion that educational attainments are related to the social capital resources available to the students. Coleman and his colleagues pointed out that social networks, norms, and expectations among community members facilitate and encourage educational achievement in the community. Coleman and his associates also attribute the existence of social networks, norms, and expectations to lowering the dropout rates of students at Catholic schools compared to dropout rates at public and other private schools (Coleman and Hoffer).

Lopez found that social capital plays a significant role in determining whether or not students enroll in college-bound curriculums. Relating social capital to the parents' involvement in the student's high school activities, Lopez concluded that students in the non college bound curriculums have lower levels of social capital at home than do those students in college-bound curriculums. Continuing, Lopez noted that students with lower levels of social capital at home also have lower levels of social capital at school. Both forms of social capital affect student performance at school.



**Table 3. Children Under 18 Years Living With Mother Only, by Marital Status of Mother, 1970 to 1995 (number in thousands)**

| Year | Total Living With One Parent | Living With Mother Only  |                          |                        |         |               |
|------|------------------------------|--------------------------|--------------------------|------------------------|---------|---------------|
|      |                              | Total Living With Mother | Marital Status of Mother |                        |         |               |
|      |                              |                          | Divorced                 | Married, Spouse Absent | Widowed | Never Married |
| 1995 | 18,938                       | 16,477                   | 6,019                    | 3,901                  | 695     | 5,862         |
| 1994 | 18,590                       | 16,334                   | 5,799                    | 3,838                  | 696     | 6,000         |
| 1993 | 17,872                       | 15,586                   | 5,687                    | 3,739                  | 649     | 5,511         |
| 1992 | 17,578                       | 15,396                   | 5,507                    | 3,790                  | 688     | 5,410         |
| 1991 | 16,624                       | 14,608                   | 5,206                    | 3,583                  | 780     | 5,040         |
| 1990 | 15,867                       | 13,874                   | 5,118                    | 3,416                  | 975     | 4,365         |
| 1989 | 15,493                       | 13,700                   | 5,227                    | 3,380                  | 803     | 4,290         |
| 1988 | 15,329                       | 13,521                   | 5,010                    | 3,371                  | 838     | 4,302         |
| 1987 | 15,071                       | 13,420                   | 5,325                    | 3,288                  | 821     | 3,985         |
| 1986 | 14,759                       | 13,180                   | 5,350                    | 3,322                  | 902     | 3,606         |
| 1985 | 14,635                       | 13,081                   | 5,280                    | 3,367                  | 939     | 3,496         |
| 1984 | 14,025                       | 12,646                   | 5,167                    | 3,423                  | 925     | 3,131         |
| 1983 | 14,006                       | 12,739                   | 5,190                    | 3,334                  | 1,004   | 3,212         |
| 1982 | 13,702                       | 12,512                   | 5,103                    | 3,518                  | 1,123   | 2,768         |
| 1981 | 12,619                       | 11,416                   | 4,912                    | 3,540                  | 1,158   | 1,807         |
| 1980 | 12,466                       | 11,406                   | 4,766                    | 3,610                  | 1,286   | 1,745         |
| 1980 | 12,162                       | 11,131                   | 4,630                    | 3,519                  | 1,260   | 1,721         |
| 1979 | 11,529                       | 10,531                   | 4,259                    | 3,487                  | 1,241   | 1,544         |
| 1978 | 11,711                       | 10,725                   | 4,335                    | 3,509                  | 1,250   | 1,633         |
| 1977 | 11,311                       | 10,419                   | 4,211                    | 3,618                  | 1,255   | 1,335         |
| 1976 | 11,121                       | 10,310                   | 4,017                    | 3,797                  | 1,357   | 1,139         |
| 1975 | 11,243                       | 10,231                   | 3,644                    | 3,857                  | 1,565   | 1,166         |
| 1974 | 10,489                       | 9,647                    | 3,278                    | 3,789                  | 1,614   | 966           |
| 1973 | 10,093                       | 9,272                    | 3,103                    | 3,745                  | 1,533   | 892           |
| 1972 | 9,634                        | 8,838                    | 2,799                    | 3,901                  | 1,506   | 632           |
| 1971 | 9,478                        | 8,714                    | 2,622                    | 3,866                  | 1,449   | 773           |
| 1970 | 8,199                        | 7,452                    | 2,296                    | 3,234                  | 1,395   | 527           |

NOTE: Data based on the Current Population Survey (CPS) unless otherwise specified.



Indicator variables selected to measure social capital associated with education used in this study include: high school graduation rates and percentages of teens not in school.

*Crime.* Increasing litigation reflects a reduced ability to resolve disputes without engaging the judicial system. In addition, increasing litigation rates suggests increased transaction costs that tend to reduce labor specialization and trade and generally lower the level of economic activity. Increasing litigation may also reflect a decrease in effective property rights that are made legitimate by their acceptance by non-owners, reflecting a decrease in social capital. States with high rates of litigation are assumed to reflect low levels of social capital. Finally, the complete breakdown of social capital is reflected by violent deaths, an increasing problem among youth.

Indicator variables selected to measure antipathy based social capital associated with crime include: litigation rates and violent death rates for teens.

*Labor force participation and poverty.* The final category reflective of social capital is the employment network. In a well-functioning and social capital rich community, the skills of workers are employed productively. Moreover, this employment network communicates job opportunities efficiently. In addition, job training opportunities are also communicated. Two separate studies confirm that employment in nearly 75% of the cases is obtained through informal contacts (Granovetter; U.S. Department of Labor). Where social capital enhanced labor networks operate efficiently, it is expected that there will exist high levels of labor force participation. Associated with labor force participation is the economic well-being of children. Thus, at the same time we measure labor force participation, we examine childhood poverty rates.

Indicator variables selected to measure social capital reflected in the labor market include labor force participation rates and childhood poverty rates.

*Transfer payments for health purposes.* Most state and federal transfer payments are included in household income measures. However, public expenditures for health maintenance are not included. It is assumed here that transfer payments for health maintenance are an important reflection of a society's collective social capital and are included as a separate variable.

## Measuring the Connection Between Social Capital and Income Distributions

We now consider methods to test empirically whether or not increases in social capital increase the level and reduce the disparity of income between households. Testing for the effect of increased social capital on the level and disparity in household income requires that we make inferences that depend on communities instead of the two-member economy for which theoretical results were derived in Appendix A.

Earlier we defined the mean levels of household social capital and income as  $\mu_s$  and  $\mu_y$ , respectively. Then, we denote the standard deviations of household social capital and income as  $\sigma_s$  and  $\sigma_y$ , respectively. The important question is: do increases in the means and standard deviation of social capital have any predictable effects on the means and standard deviations of household income? Our maintained hypotheses based on our earlier deductions are the following:

$$H_o : \frac{d\mu_y}{d\mu_s} > 0 \quad (10)$$

and:

$$H_o : \frac{d\sigma_y}{d\sigma_s} > 0 \quad (11)$$

We would like to test directly the relationships between social capital and household income hypothesized in equations (10) and (11). However, we are not able to observe social capital directly. Instead, what we observe are indicator



variables already described that are expected to be highly correlated with social capital. These indicator variables can then be tested to see how they correlate with the averages, standard deviations, and CVs of household income.

## **IX. Testing the Relationship Between Indicator Variables and the Level and Disparity of Household Income**

### **Robustness of Results**

In strong statistical tests there is a small likelihood that the tests will confirm an incorrect hypothesis. Our effort to apply this statistical requirement to our study takes the following form. We perform the same analysis on two different data sets.<sup>10</sup> The assumption is that there is only a small likelihood that an incorrect hypothesis would be confirmed two times using two different data sets. The first data set is organized by states, race, and ethnic origin for 1990. The second data set is organized by states, race, and ethnic origin for 1980.

Ex ante, we expect strong correlations between indicator variables within each group. Second, we expect to find strong correlations between indicator variables and income distribution measures including means of household income and CVs of household income. Finally, using a statistical procedure to capture the influence of our indicator variables, factor analysis, we intend to determine the extent to which means of household income and CVs of income for households can be predicted. Finally, we expect to find an inverse relationship between means of household income and CVs of household income.

### **Sources of Data**

The primary data sources for this study are the U.S. Bureau of the Census reports, Statistical Abstract of the United States, the Economic Report to the President, and other publications. Using Public Use Microdata Sample (PUMS) data, we calculated means and CVs for household

income in states for census years 1980 and 1990 (reported for 1979 and 1989, respectively).

Sources of data and values of indicator variables used in this report are included in a supplement to this report. The supplement serves two purposes. The first purpose of the supplement is to make available the data used in this report to other researchers who wish to replicate our results or to perform additional tests. The second purpose of the supplement is to reduce the length of this report. The title of the supplement to this report is: *Data Book: Social Capital and Household Income Distributions*. Throughout the remainder of this study, instead of listing data sources, the reader will be referred to the *Data Book*.

Because of the large number of variables included in this analysis, there is a need to standardize their names. To assist the reader in identifying variables described later on, the acronyms used to describe variables are now identified. Each variable's acronym contains a prefix and a suffix. The prefix identifies the variable group to which the variable belongs and the year represented by the data. The group designations are F for family, E for education, C for crime, T for transfer payments, and L for labor. If the variable is computed, the variable prefix is Cu. Also included in the prefix are the year designations, either 90 for year 1990 or 80 for year 1980. Thus, the prefix for a variable from the family group in 1980 would have a prefix F80.

The suffix identifiers for variables used in this study are letters or acronyms used to designate the variable's name. The variables and the identifying acronym in parentheses include: percentages of households headed by a single female with children (HHSFC), birth rates of single teens (BRST), infant mortality rates (IMR), high school graduation rates (HSGR), percentages of teens not in school (TNIS), litigation rates (LIT), violent death rates for teens (VDT), labor force participation rates (LFPR), and childhood poverty rates (CPR). The computed variables in this study are means of household income (M), and CVs of household income. Transfer payment variables



are health maintenance expenditures (H), welfare expenditures (W), and education expenditures (E).

Examples of complete variable designations including prefixes and suffixes follow. The variable representing litigation rates in 1990 is represented as C90/LIT. The mean of household income in 1980 is described as Cu80/M. State expenditures per student in 1990 in support of public education are represented as T90/E.

## **X. Statistical Results for States, 1990**

### **Correlations Between Indicator Variables**

Correlations and statistical significance levels between indicator variables for states in 1990 are reported in Table 4. Correlations between indicator variables within groups are bolded and boxed. All of the correlations within variable groups are significant at the .1% level or higher. For variables included in the 1990 family integrity variable group, (F90), the correlation between the percentage of households headed by a single female with children (F90/HHSFC), and birth rates of single teens (F90/BRST) is 67%. The correlation between F90/HHSFC and infant mortality rates (F90/IMR) is 59%. The correlation between F90/BRST and F90/IMR is 75%.<sup>11</sup>

For variables included in the 1990 education variable group (E90), the correlation between high school graduation rates (E90/HSGR) and percentages of teens not in school (E90/TNIS) is -61%.

For variables included in the 1990 crime variable group (C90), the correlation between litigation rates measured as per capita civil cases (C90/LIT) and violent death rates for teens (C90/VDT) is 44%.

Finally, for variables included in the 1990 labor force participation group (L90), the correlation between labor force participation rates (L90/LFPR) and child poverty rates (L90/CPR) is -70%.

It is also significant that F90/HHSC is correlated with all of the other indicator variables at the .1% or higher except for the C90/VDT.

To measure the correlations between indicator variables and income distribution measures suggested by equations (10) and (11) requires measures of means and standard deviations of household income. Using data from the 1990 U.S. Census, we estimated the means and standard deviations of household income by state.<sup>12</sup> However, the means and standard deviations between states may not be directly comparable if there exist significant differences in the cost of living between states.

To convert income distributions to the same index would require that we calculate cost of living indices for each state. However, a consistent measure of the dispersion of income could be obtained by dividing the unadjusted standard deviation of income by the unadjusted mean income measure to obtain CVs. CVs provide a consistent measure of the average dispersion as a percent of the mean. But the problem remains: if significant differences in the cost of living exist between states, how do we scale the differences?

This study did make an effort to obtain cost of living indices by states and the adjustment is described in the *Data Book*. But we choose to use the unadjusted data because arbitrage, especially between major population centers, was expected to reduce significant differences in the cost of living. In addition, after adjusting mean data, we were disappointed with the statistical properties of the adjusted means and used instead unadjusted means.

Having measures of means and CVs of household income by state, we then calculated the correlations between indicator variables and CVs and means of household income. These correlations are reported in Tables C1 and C2, respectively.



**Table 4. Correlations and Significance Levels Between Indicator Variables Representing Social Capital Associated with Family Integrity, Educational Achievements, Crime, and Labor Market Participation by States, 1990<sup>a</sup>**

|   | Correlation Coefficients and Significance Levels |              |                 |               |             |              |             |               |
|---|--|--------------|-----------------|---------------|-------------|--------------|-------------|---------------|
|   | Family (F90)                                     |              | Education (E90) |               | Crime (C90) |              | Labor (L90) |               |
|   | BRST   | IMR          | HSGR            | TNIS          | LIT         | VDT          | LFPR        | CPR           |
| Family (F90)  |  |              |                 |               |             |              |             |               |
| HHSFC<br>(Percentages of<br>Households Headed by a<br>Single Female with<br>Children) | <b>.6743</b>                                     | <b>.5936</b> | -.6562          | .5116         | .4704       | .2115        | -.4017      | .6584         |
|   | <b>.000</b>                                      | <b>.000</b>  | .000            | .000          | .000        | .140         | .004        | .000          |
| BRST<br>(Birth Rates of Single<br>Teens)  |  | <b>.7468</b> | -.5398          | .5125         | .4829       | .5235        | -.5743      | .7638         |
|   |  | <b>.000</b>  | .000            | .000          | .000        | .000         | .000        | .000          |
| IMR<br>(Infant Mortality Rates)   |  |              | -.4232          | .3556         | .4876       | .3800        | -.2924      | .4802         |
|   |  |              | .002            | .011          | .000        | .006         | .039        | .000          |
| Education (E90)   |  |              |                 |               |             |              |             |               |
| HSGR<br>(High School Graduation<br>Rates)   |  |              |                 | <b>-.6071</b> | -.3166      | -.2498       | .2371       | -.5661        |
|   |  |              |                 | <b>.000</b>   | .025        | .080         | .097        | .000          |
| TNIS<br>(Percentage of Teens Not<br>in School)  |  |              |                 |               | .4996       | .3882        | -.3832      | .6059         |
|   |  |              |                 |               | .000        | .005         | .006        | .000          |
| Crime (C90)   |  |              |                 |               |             |              |             |               |
| CRM<br>(Litigation Rates)   |  |              |                 |               |             | <b>.4380</b> | -.3816      | .4842         |
|   |  |              |                 |               |             | <b>.001</b>  | .006        | .000          |
| VDT<br>(Violent Death Rates for<br>Teens)   |  |              |                 |               |             |              | -.2552      | .5298         |
|   |  |              |                 |               |             |              | .074        | .000          |
| Labor (L90)   |  |              |                 |               |             |              |             |               |
| LFPR<br>(Labor Force Participation<br>Rates)  |  |              |                 |               |             |              |             | <b>-.6998</b> |
|   |  |              |                 |               |             |              |             | <b>.000</b>   |

Source: Estimated by the authors.

<sup>a</sup> The significance level of the second number is the probability that the correlation between variables is zero.



The correlation between CVs calculated by state for 1990 (Cu90/CV) and the indicator variables all have the predicted sign and all are significant. The most significant correlations are between Cu90/CV and L90/CPR (76%), L90/LFPR (-68%), and F90/BRST (54%).

Correlations between the means of household income and the indicator variables are less significant than were the correlations between the indicator variables and the CVs. These results may suggest that social capital has more to do with income disparity than it does with the level of income. Or it may mean that a proper cost of living index is needed to fully reflect the appropriate values of the data.

Four of the indicator variables were significantly correlated with the means of household income and all of the four carried expected signs. Significantly correlated with the means of household income in 1990 (Cu90/M) were L90/LFPR (58%), L90/CPR (46%), C90/VDT (-44%), and F90/BRST (-41%). The five indicator variables not significantly correlated with Cu90/M were: F90/HHSFC, F90/IMR, E90/HSGR, E90/TNIS, and C90/LIT.

### Factor Analysis

Because of the large number of indicator variables, their influence was summarized using factor analysis.<sup>13</sup> Eighty-four percent of the variance associated with the indicator variables was captured using four factors. The factors are listed in Table C3. The factors are consistent with the family, education, labor, and crime groupings described earlier.

Next, efforts were made to predict differences in means and CVs in states during 1990 using the four factors described in Table C3 and one income transfer variable. The one income transfer variable used was per person expenditures by states for health. Regression results using the four factors and the one transfer variable to predict CVs and means of household income for states in 1990 are reported in Tables C4 and C5, respectively.

In the regression equation used to predict Cu90/CV, the labor and crime factors were significant at .01% level. The education factor was significant at the 2% level of significance and the family factor was significant at the 9% level. The expenditures per person for health costs variable was significant in reducing the CVs at the 6% level of significance.

In the regression equation used to predict means of household income by state, the education factor was significant at the 10% level. The family factor was not significant. The labor and crime factors were significant at the .1% level or higher. Finally, transfer payments for health were significant at the .7% level.

### Social Welfare Functions and Preferred Income Distributions, 1990

Besides the social capital hypotheses that relate changes in social capital to changes in the level and dispersion of income, another important question is: can we assert that one household income distribution is preferred to another? The answer is yes if some important assumptions are adopted.

Assume that all individuals derive the same utility from their own consumption of income and that selfishness of preference dominates. Assume also that in the society under investigation that each member's utility function measured over own income is increasing and concave down. Furthermore, assume that each income distribution is related to another by location-scale (Meyer).

If, in addition to the assumptions made above, the utility functions satisfy von Neumann-Morgenstern axioms underlying expected utility, then it is possible to rank distributions using a mean-standard deviation frontier. This ranking suggests that if two household income distributions have equal means but different standard deviations of household income, the distribution with the smaller standard deviation of income is socially preferred because it generates a greater level of satisfaction for society. On the other hand, if one distribution has a higher mean



for household income and also a greater standard deviation, then society cannot indicate a preference for one distribution or the other without imposing much stronger restrictions on each member of society's utility function.<sup>14</sup> The mean-standard deviation frontier for states in 1990 is presented in Figure 8.

If we are willing to add additional assumptions about the slopes of indifference curves in mean-standard deviation space, then we may find the preferred income distributions along a mean-CV frontier. Along the mean-CV frontier, distributions with the highest means for the same CV are preferred. Whether the slope along the mean-CV frontier is positive or negative depends on the slope of the mean-standard deviation frontier. If the slope on the mean-standard deviation frontier exceeds one, that is, mean values are increasing faster than standard deviations, the slope of the mean-CV frontier is downward sloping.

The mean-CV frontier for states in the 1990 frontier is described in Figure 9 and is downward sloping consistent with our earlier deductions. According to our criteria, those states with the highest means for household income in Figure 9 dominate since they also have the lowest CVs.

CVs and means of household income in descending order are reported in Tables C6 and C7. The inference from Figure 9 and Tables C6 and C7 is that states differ in their social capital bases and some states have income distributions preferred to those in other states. Maryland, Hawaii, Utah, Washington, and Virginia have the lowest CVs. Connecticut, New Jersey, Maryland, and California have the highest means of household income.

Finally, Table C8 reports the regression results of means regressed on CVs associated with household income for states in 1990. The regression results support relationships described in Figure 5 that means of household income decrease with increases in CVs.

## **XI. Statistical Results for States, 1980**

### **Correlations Between Indicator Variables**

Correlations and statistical significance levels for indicator variables for states in 1980 are reported in Table 5. Correlations between indicator variables within groups are bolded and boxed as they were in Table 4. Of particular interest for this study is to compare the correlations between 1980 and 1990 for states. Robustness of results would lead us to predict similar results between 1980 and 1990. Any significant changes between 1980 and 1990 should be explained in terms of changing levels of social capital.

The most significant change in the 1980 and 1990 correlations was between percentages of households headed by a single female with children and birth rates of single teens. In 1990, the correlation was 67%. In 1980, the same correlation was only 8% and not significant. The change in the correlations between percentages of households headed by a single female with children and birth rates of single teens variables reflects an important national trend. The trend is the increasing rate of households headed by a single female with children who have never married.

To measure the correlations between indicator variables and income distribution measures suggested by equations (10) and (11), we again calculated means and CVs using the 1980 U.S. Bureau of the Census household income data. For the reasons described earlier, we measured income dispersion using CVs.

Having measures of means and CVs of household income by state for 1980, we next calculated the correlations between indicator variables and CVs and means of household income. These correlations are reported in Tables D1 and D2, respectively.



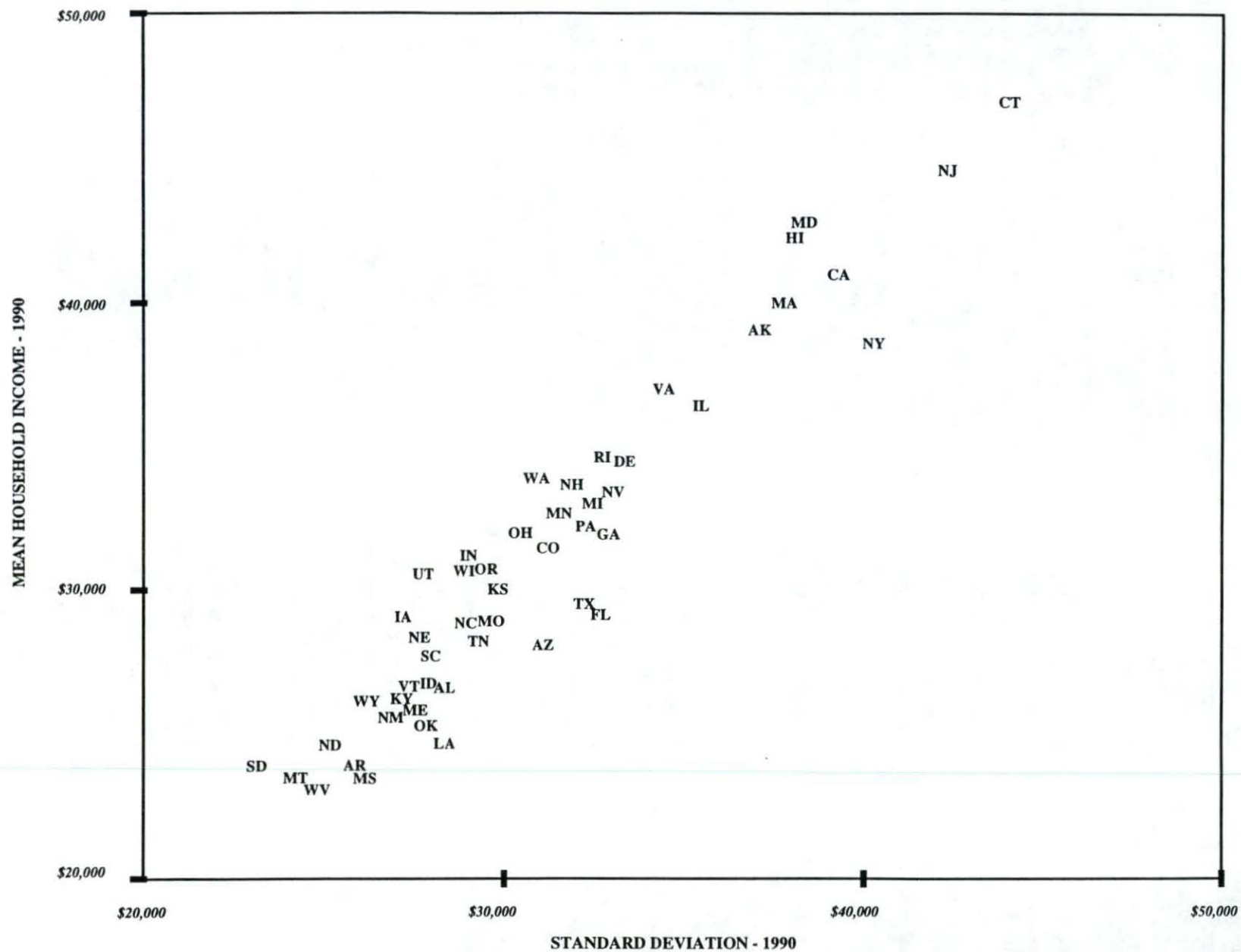


Figure 8. The Mean-Standard Deviation Frontier for States in 1990



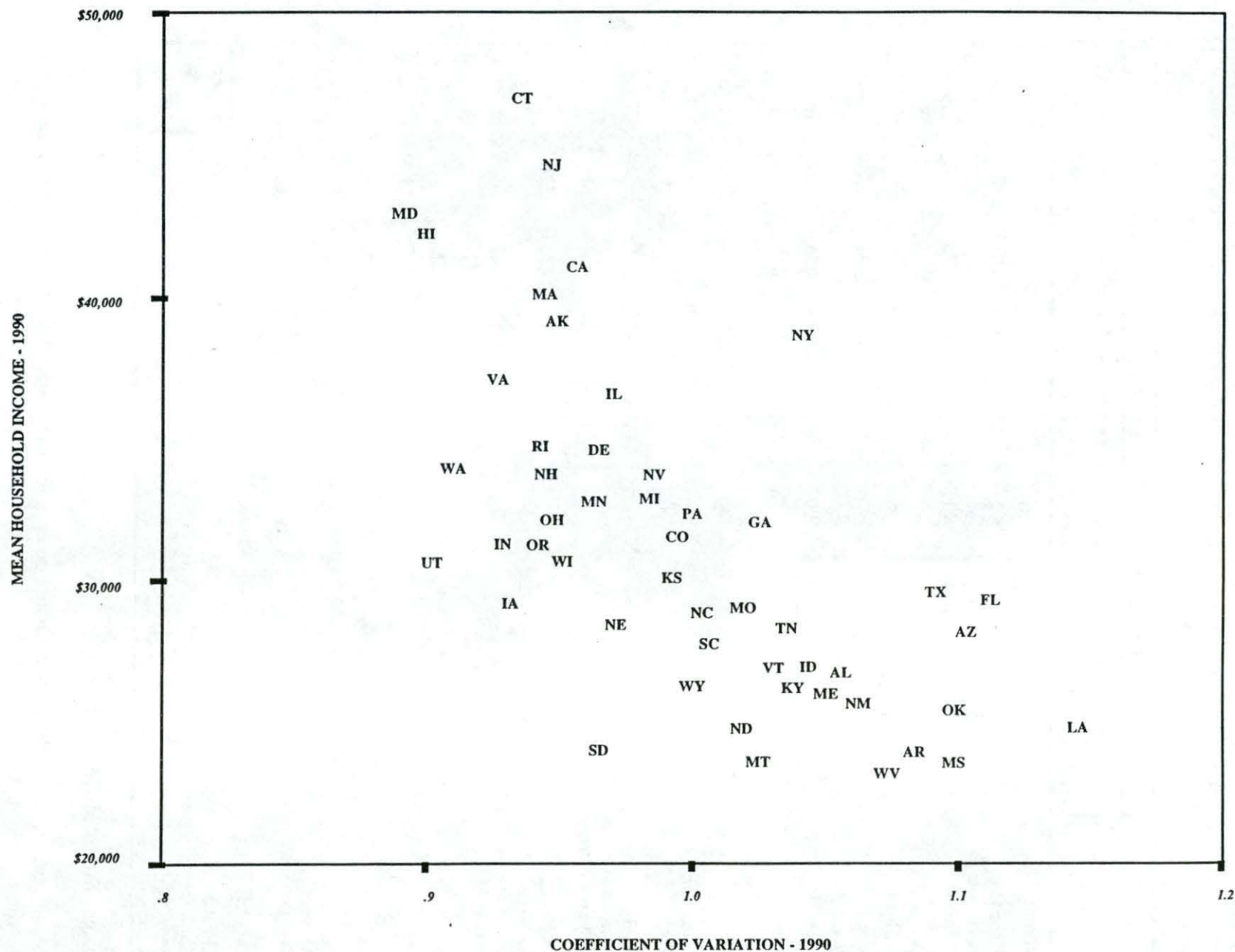


Figure 9. The Mean-Coefficient of Variation Frontier for States in 1990



**Table 5. Correlations and Significance Levels Between Indicator Variables Representing Social Capital Associated with Family Integrity, Educational Achievements, Crime, and Labor Market Participation by States, 1980<sup>a</sup>**

|   | Correlation Coefficients and Significance Levels |                |               |                 |                |                |               |                |                |  |
|---|--|----------------|---------------|-----------------|----------------|----------------|---------------|----------------|----------------|--|
|   | Family (F80)                                     |                |               | Education (E80) |                | Crime (C80)    |               | Labor (L80)    |                |  |
|   | HHSFC  | BRST           | IMR           | HSGR            | TNIS           | LIT            | VDT           | LFPR           | CPR            |  |
| Family (F80)  |  |                |               |                 |                |                |               |                |                |  |
| HHSFC<br>(Percentages of Households Headed by Single Females with Children) |  | .0825<br>.5696 | .3487<br>.013 | .0054<br>.971   | .1814<br>.207  | .4305<br>.002  | .2533<br>.076 | .2568<br>.072  | .1869<br>.194  |  |
| BRST<br>(Birth Rates of Single Teens)                                       |  |                | .6623<br>.000 | -.7709<br>.000  | .7689<br>.000  | .4815<br>.000  | .2718<br>.056 | -.5554<br>.000 | .7841<br>.000  |  |
| IMR<br>(Infant Mortality Rates)   |  |                |               | -.6684<br>.000  | .4119<br>.003  | .4484<br>.001  | .0682<br>.638 | -.3981<br>.004 | .6779<br>.000  |  |
| Education (E80)   |  |                |               |                 |                |                |               |                |                |  |
| HSGR<br>(High School Graduation Rates)                                      |  |                |               |                 | -.4618<br>.001 | -.2915<br>.040 | .1505<br>.297 | .6264<br>.000  | -.7777<br>.000 |  |
| TNIS<br>(Percentage of Teens Not in School)                                 |  |                |               |                 |                | .4187<br>.002  | .4321<br>.002 | -.3268<br>.021 | .5529<br>.000  |  |
| Crime (C80)   |  |                |               |                 |                |                |               |                |                |  |
| LIT<br>(Litigation Rates)   |  |                |               |                 |                |                | .4343<br>.002 | -.2011<br>.161 | .4466<br>.001  |  |
| VDT<br>(Violent Death Rates for Teens)                                      |  |                |               |                 |                |                |               | .0729<br>.615  | .1949<br>.175  |  |
| Labor (L80)   |  |                |               |                 |                |                |               |                |                |  |
| LFPR<br>(Labor Force Participation Rates)                                   |  |                |               |                 |                |                |               |                | -.6555<br>.000 |  |

Source: Estimated by the authors.

<sup>a</sup> Significance levels of the second number on each level represent the probability that the correlation between variables is zero.



The correlations between CVs (Cu80/CV) and the indicator variables are all significant and have the correct sign except for F80/HHSFC and C80/VDT which are not significant. The most significant correlations are between Cu80/CV and F80/BRST (70% compared to 54% in 1990), L80/CPR (86% compared to 76% in 1990), E80/HSGR (-72% compared to -34% in 1990), E90/TNIS (51% compared to 39% in 1990), and L80/LFPR (-67% compared to -68% in 1980).

Correlations between means of household income (Cu80/M) and indicator variables reported in Table D2 showed three of the indicator variables to be insignificant. These three variables uncorrelated with Cu80/M were F80/IMR, C80/LIT, and C80/VDT. The remaining indicator variables were significant and had the expected signs except for F80/HHSFC which was unexpectedly positively correlated with Cu80/M. This coupled with the insignificant correlation between F80/HHSFC and Cu80/CV suggests what the variable HHSFC represents has changed between 1980 and 1990. This change that has been referred to earlier is the increasing importance of births from unwed mothers in the creation of households headed by a single female with children.

### Factor Analysis

We intended to determine the extent to which CVs and means of household income could be predicted using our indicator variables. Because of the large number of indicator variables, their influence was summarized using factor analysis. Eighty-seven percent of the variance associated with the indicator variables was captured using four factors. The factors are listed in Table D3. Two of the factors represent combinations of variables from the four indicator variable groups. The remaining two factors contain family integrity variables and crime variables.

Next, efforts were made to predict differences in 80/CV and 80/M using the four factors described in Table D3 and one income transfer variable, 80/H. Regression results using the four factors and one transfer variable to predict 80/CV and 80/M are reported in Tables D4 and D5,

respectively. In each regression, all factors but one were significant.

In the regression equation used to predict Cu80/CV, all but the family factor were significant at less than the 1% level of significance. In the regression equation used to predict means of household income, all but the crime factor were significant at less than the 1% level of significance. In addition, T80/H variable was not significant in increasing Cu80/M.

CVs and means of household income in descending order are reported in Tables D6 and D7. Mean and standard deviations and means and CVs are described graphically in Figures 10 and 11, respectively. The inference from Figures 10 and 11 and Tables D6 and D7 is that states differ in their social capital bases and some states have income distributions preferred to those in other states. Alaska, Wyoming, Utah, and New Hampshire have the lowest CVs. Alaska, Hawaii, Maryland, Connecticut, and New Jersey have the highest means of household income.

### Social Welfare Functions and Preferred Income Distributions, 1980

Figure 11 describes the relationships between means of household income for states in 1980 (Cu80/M) and CVs for states in 1980 (Cu80/CV). The relationship between Cu80/M and Cu80/CV is negative in Figure 11 as it was in Figure 9. This negative relationship points out the tendency for Cu80/M to increase at a faster rate than does the standard deviation of household income. Thus, CVs increase with decreases in the means of household income. As a result, if one uses Cu80/CV as one's measure of income inequality as we do, then there is no reduction in income disparity associated with reduced values of Cu80/M.

Finally, Table D8 reports the regression results of means of household income on CVs measured by state. The regression results support the view that means of household income decrease with increases in CVs.



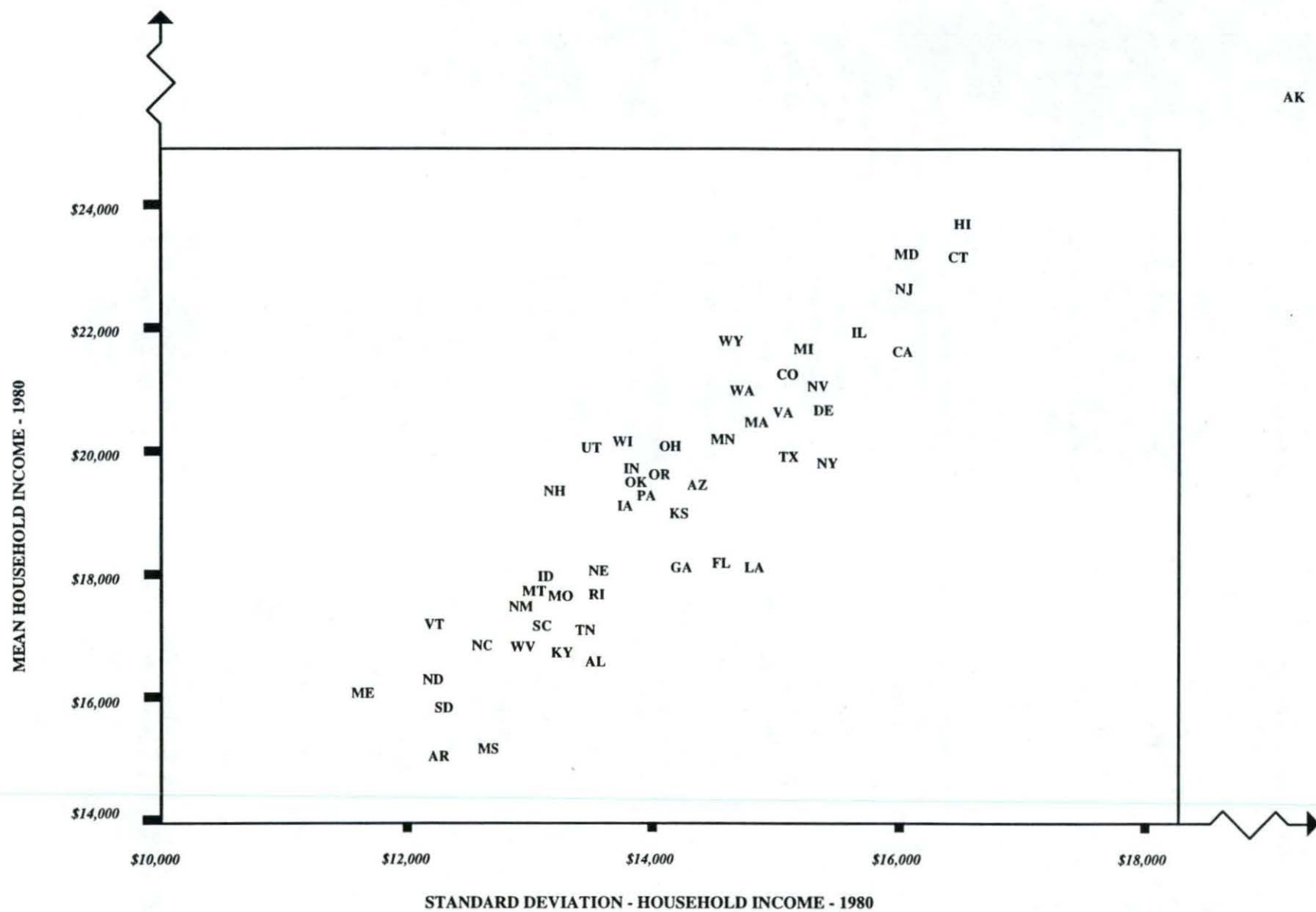


Figure 10. The Mean-Standard Deviation Frontier for States in 1980



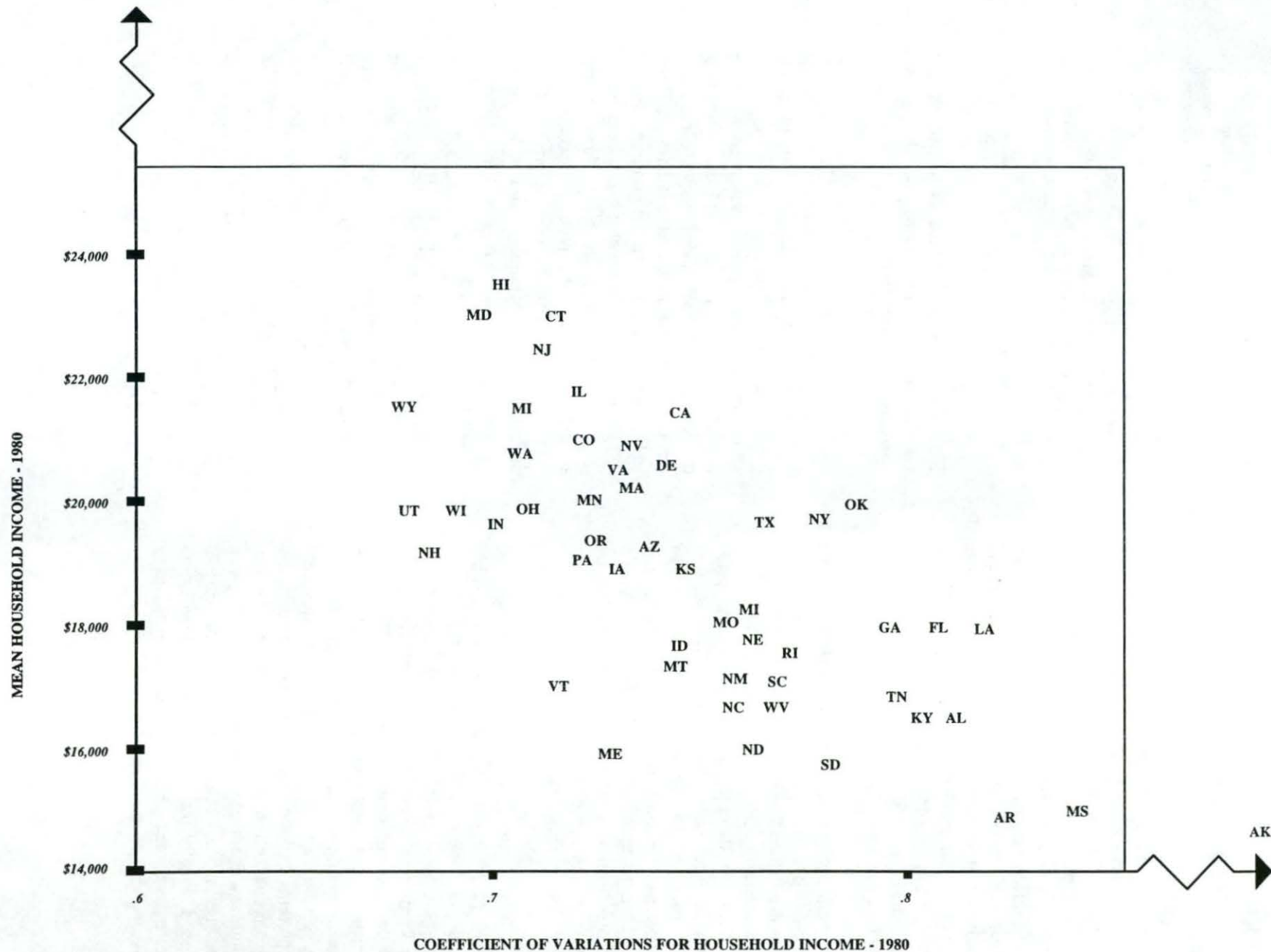


Figure 11. The Mean Coefficient of Variation Frontier for States in 1980



## XII. A Comparison of 1980 and 1990 Income Distributions

Having calculated means and CVs of household income by state for 1980 and 1990, it was next determined to construct a measure of the changes in these variables between 1980 and 1990. One measure is obtained by forming the ratio of means of household income in 1990 and 1980 after adjusting 1980 dollars to their equivalent 1990 dollars. A ratio of one would indicate that the means of household income have not changed in real terms. A ratio of greater than (less than) would suggest real incomes for households have improved (become worse) on average between 1980 and 1990.

Similarly, ratios were formed from CVs calculated in 1990 and 1980. Dividing CVs calculated for 1990 by CVs calculated for 1980 forms a ratio of two percentages which do not require adjustments for changes in the price indices. A ratio of  $Cu90/CV$  to  $Cu80/CV$  equal to one suggests that on average, the percentage dispersion of income around the mean income has not changed during the decade of the 80's. A ratio greater than (less than) one indicates that on average, the dispersion of income around the mean has increased (decreased).

To evaluate the changes between 1980 and 1990, one might consider a quadrant of possible values. Let the vertical axis represent the ratios of means of household income, and let the horizontal axis represent the ratios of CVs. Then, the center of the axis represents a stationary position. The northeast quadrant represents increases in the mean ratio, a good, but also increases in dispersion, a bad. The northwest quadrant represents increases in the mean ratio, a good, and also reductions in dispersion, also a good. The southwest quadrant represents decreases in mean incomes, a bad, but also decreases in dispersion, a good. Finally, the southeast quadrant represents decreases in mean income, a bad, and increases in dispersion, also a bad (see Figure 12).

|                         |  |  |
|-------------------------|--|--|
| $\frac{Cu90/M}{Cu80/M}$ | Increased Real Income and Reduced Disparity of Income (Goods)  | Increased Real Income and Increased Disparity of Income (Good/Bad) |
| 1                       | Reduced Real Income and Reduced Disparity of Income (Bad/Good) | Reduced Real Income and Increased Disparity of Income (Bads)       |
|                         | 1  | $\frac{Cu90/CV}{Cu80/CV}$  |

**Figure 12. Relative Changes in Means and Coefficients of Variations of Household Incomes, 1980 and 1990**

Figure 13 plots the actual ratios of means and CVs by states. It would be preferable to see the observations located in the northwest quadrant. Unfortunately, there were no observations in the northwest quadrant, nor in the southwest quadrant. All of the observations were located in the northeast and southeast quadrants. Those states located in the northeast quadrant showed improved incomes but increased dispersions of incomes. Those states located in the southeast quadrant showed both reduced incomes and increased dispersion. Included in this least desirable quadrant were the states of Wyoming, Arkansas, Montana, Louisiana, West Virginia, Oklahoma, and Arizona.

In contrast, Rhode Island ranked near the top for improvement in income with the smallest increase in dispersion of income. Other states showing a significant improvement in real incomes were Connecticut, Maine, New Jersey, New York, California, and Maryland. Table E1 compares CVs and means of household income by states for 1980 and 1990. Ratios of 1990 and 1980 CVs and means of household income reflect percentage changes in the level and disparity of income during the period 1980 to 1990.

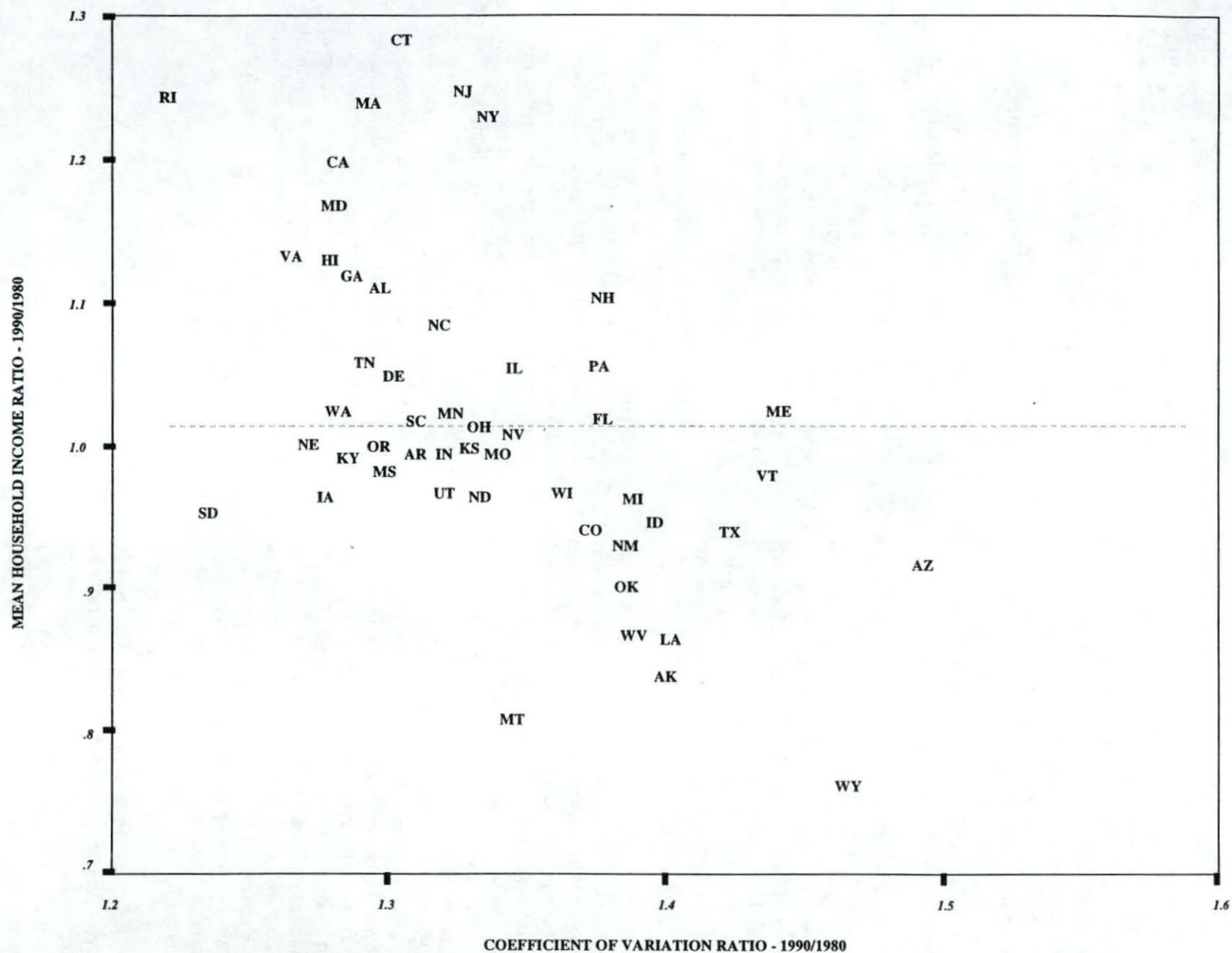


Figure 13. The Mean Ratio and Coefficient of Variation Ratio for States in 1990/1980



### **XIII. Income Distribution Changes by Race/Ethnicity**

A characteristic of today's world is the division of social capital along racial and ethnic boundaries. Efforts to reduce the consequences of social capital ordered by races and ethnicity have been discussed elsewhere and are not the focus of this study. Our interest is in examining the income distributions of racial and ethnic classes to determine the level of existing income distributions and changes in these distributions between 1980 and 1990.

It may be impossible to identify a particular individual as a member of a unique racial or ethnic group. Any one individual may belong to several depending on his or her genealogy. Nevertheless, respondents did self-identify themselves as belonging to a particular race/ethnicity and those data were used in this study to calculate means and CVs associated with household income.

Racial/ethnicity categories examined in this study were Native Americans, Asians, African Americans, Hispanics, and Whites. In addition, a total population statistic was calculated. For each race/ethnicity and for the total population, means and CVs were calculated using U.S. Census data for the two census years 1980 and 1990 and graphed in Figures 14 and 15 respectively. In 1980, Asians and Whites had the highest means of household income and also the lowest CVs. Native Americans and African Americans had nearly equal means of household income in 1980 but African Americans had a higher CV, .86 compared to .80 for Native Americans. By 1990, Asians enjoyed both the highest mean income and also the lowest CV among racial/ethnic groups. Whites continued to enjoy income levels above American Indians, Hispanics, and African Americans.

Relative changes in the income position of the races/ethnic groups between 1980 and 1990 are described in Figure 16. Native Americans showed the greatest relative improvement in levels of income between 1980 and 1990 followed by

Asians. African Americans and Hispanics saw their real income decrease between the two census years. Most significant among the changes was the disparity among whites. Compared to 1980, the disparity of income among whites increased by over 8% and registered the greatest increase in disparity for all groups. The mean ratios and ratios of CVs are reported in Table E2.<sup>15</sup>

### **XIV. Summary and Conclusions**

This paper began by presenting evidence that relationships alter the terms of trade in predictable ways. Next, the concept of social capital was introduced and defined. The similarities between social capital and other forms of capital made it possible to include its effect in a neoclassical model of utility maximization from which theoretical linkages were made between increases in social capital and changes in the mean and the disparity of income.

It was demonstrated that when an economic agent's activity produced an externality, increases in social capital would tend to internalize the externality to the agent. As a result, increases in social capital would increase the mean level of income and under some conditions reduce the difference in income. The theoretical results from the externality models also suggested that externally imposed redistributive efforts are likely to be at least partially offset by voluntary income redistribution effects realized through production decisions. However, this conclusion was deduced in a very restricted model and needs additional examination in other settings. Nevertheless, the result raises the question whether public effort at redistribution of income can be successful without sufficient levels of social capital being provided by those whose income is being transferred to lower income groups.

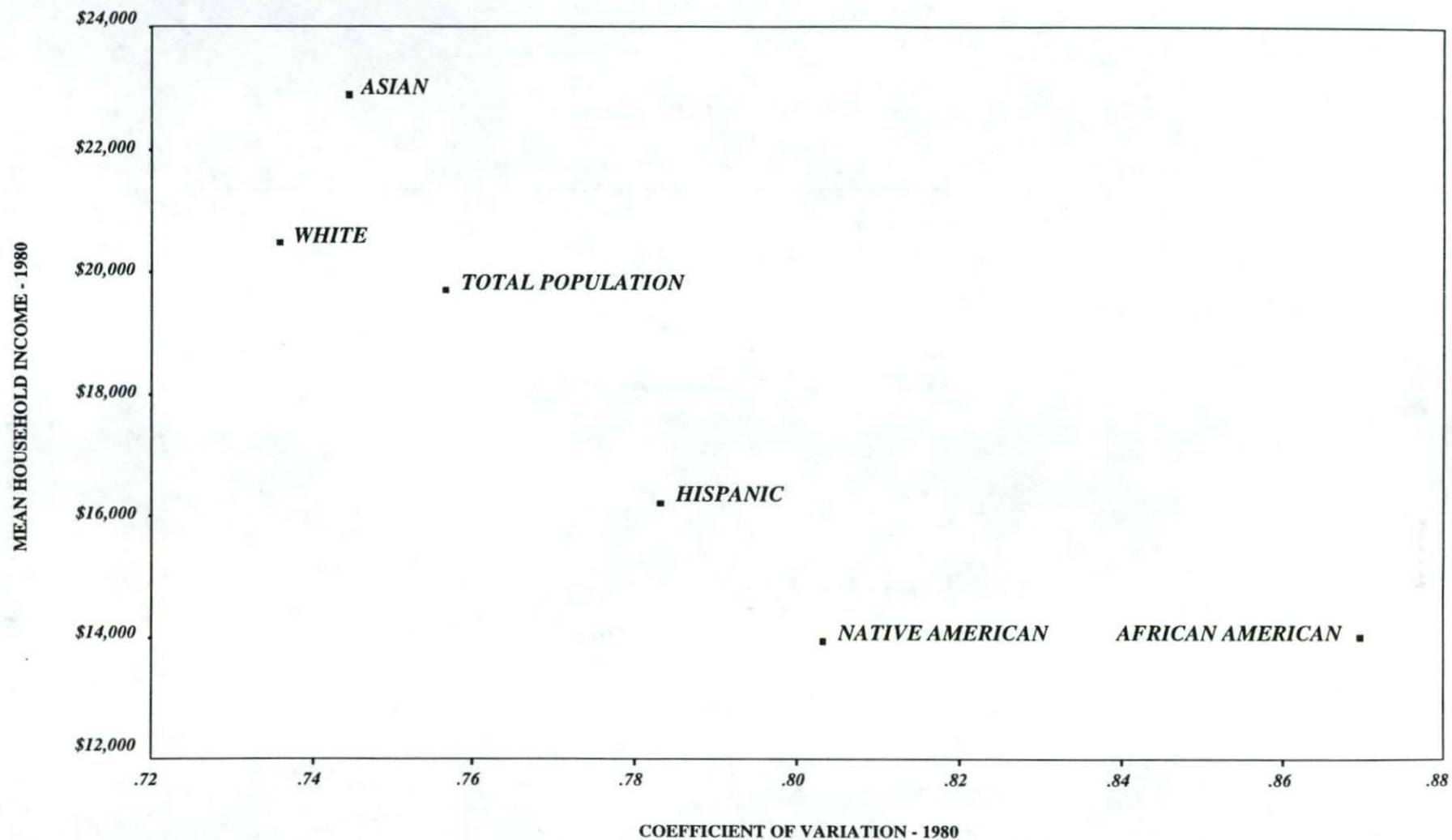


Figure 14. The Mean Coefficient of Variation Frontier by Race in 1980



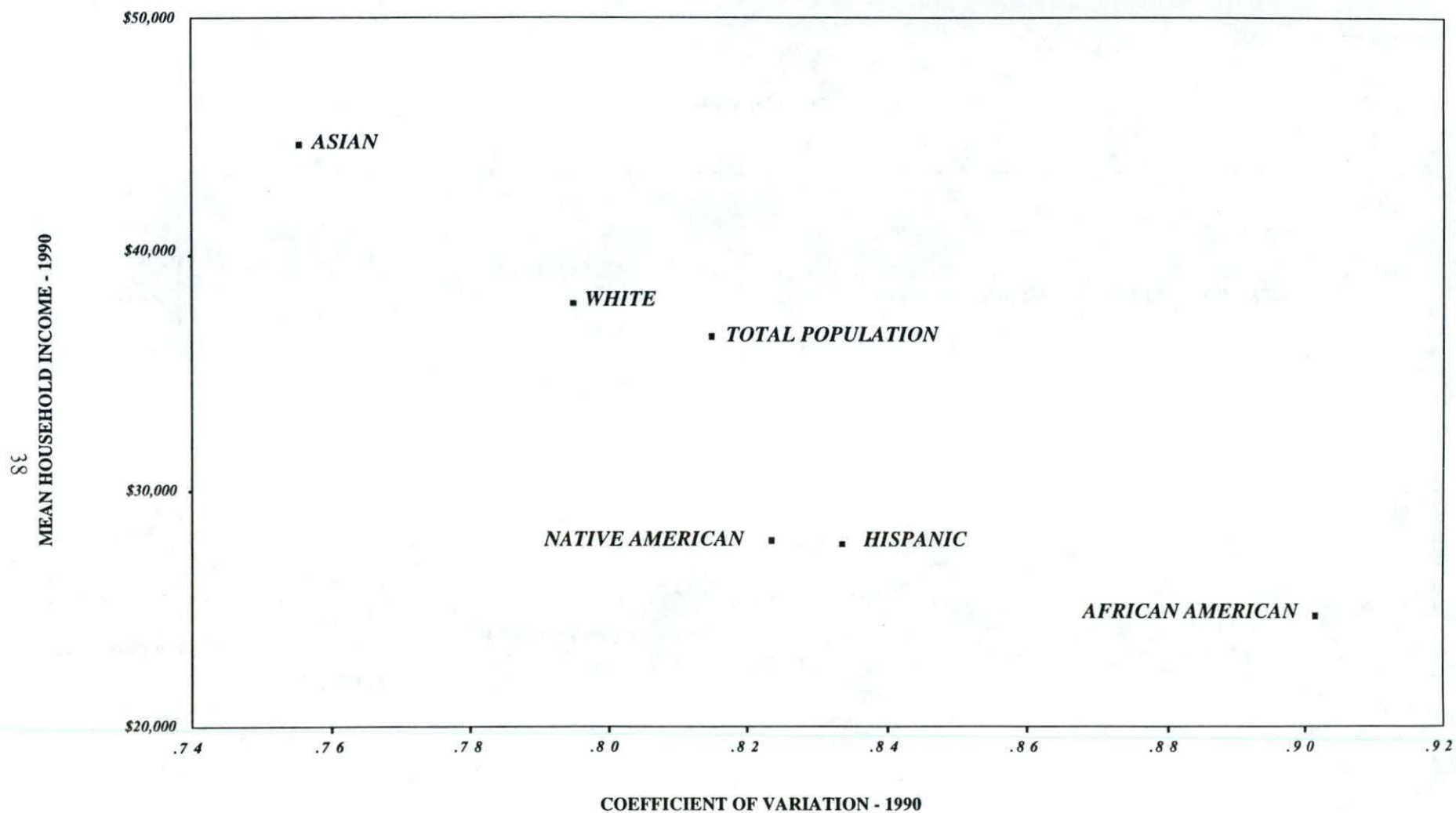


Figure 15. The Mean Coefficient of Variation Frontier by Race in 1990

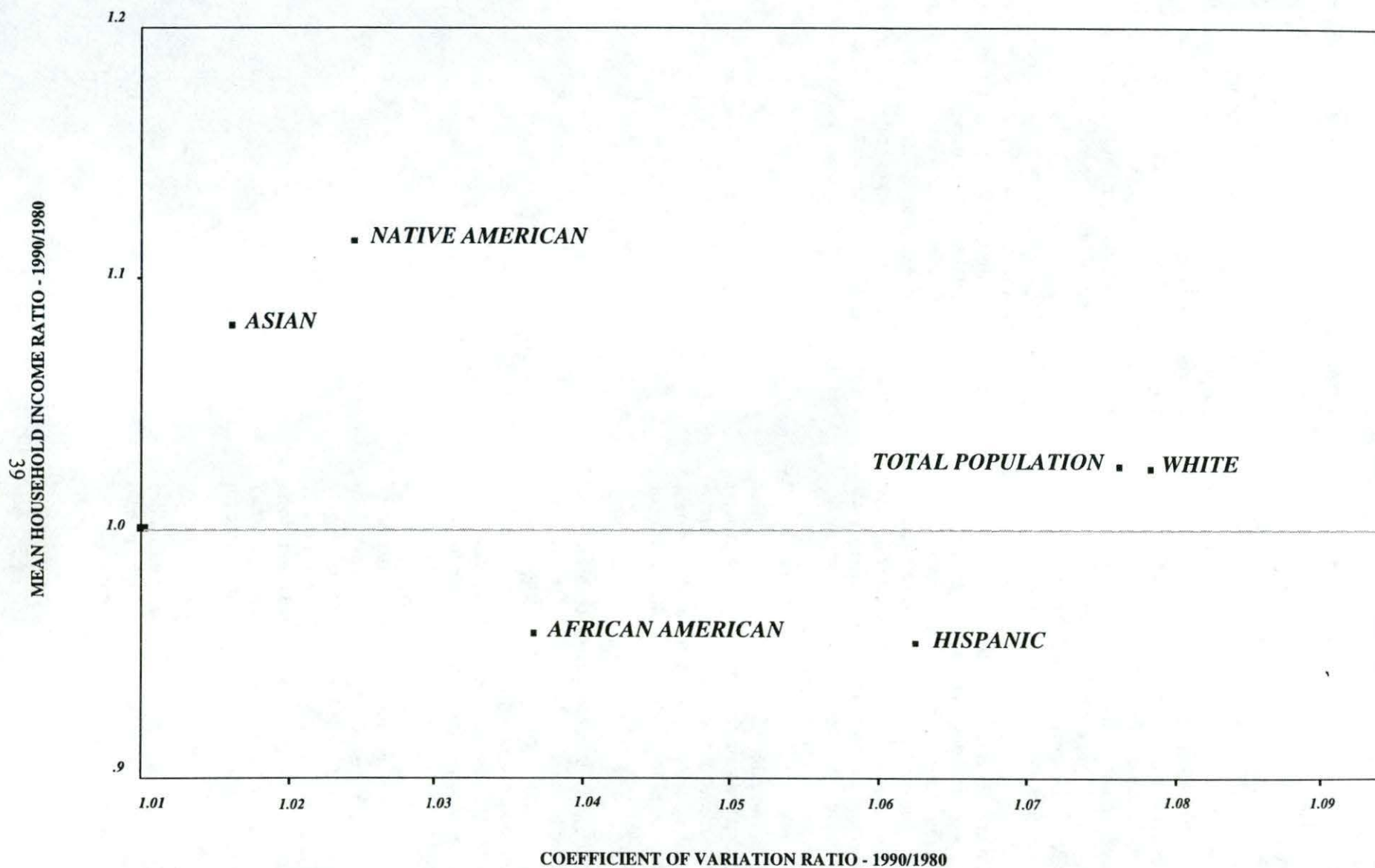


Figure 16. The Mean Income Ratio Coefficient of Variation Ratio Frontier by Race in 1990/1980



Social capital and group size were discussed next. The conclusion deduced was that opportunities for labor specialization and trade can be significantly reduced when, for whatever reason, groups divide and develop antipathetic relationships. In contrast, as the number of members of a social capital rich group increases, trade and specialization increase within the group and externalities are internalized over an increasing number of economic agents. The above results suggest that average income increases as membership in social capital rich groups increases.

Empirically relevant to the conclusion that divisions or breakdowns in existing groups have undesirable consequences on the distribution of income is the breakdown of the two-parent family in the United States. The predicted consequences on household income in the United States were a reduction in the level and an increase in the disparity of household income.

A considerable effort was then made to test the general and specific hypotheses. The general hypothesis was that decreases in social capital reflected by selected indicator variables would both reduce the level and increase the disparity of household income.

Indicator variables suggested for this purpose measured family integrity, educational achievements, crime, and labor force participation. Using primary data from the 1980 and 1990 U.S. Census and secondary data from various sources, empirical tests were performed and reported at the state level.

Mean incomes and standard deviations of income were generally positively related. But standard deviations of income were not highly correlated with our indicator variables--possibly because of scaling difficulties. Mean incomes and CVs were inversely related and generally correlated with social capital indicator variables in predicted ways.

The empirical section of this paper also tested for changes in the level and disparity of household income between 1980 and 1990 and tested for

differences in household income distributions by race/ethnicity.

The empirical results provided support for the hypothesis that changes in the indicator variables suggesting reduced levels of social capital had the effect of reducing the level of income and increasing the disparity of income between households. Interestingly enough, the indicator variables were highly correlated suggesting they were measuring a similar concept.

Obviously, the results presented in this paper need to be tested in other settings. Such analyses might examine the relationship between increases in social capital and the means and CVs in counties, cities, and businesses. Still, other studies might examine in more detail which factors contribute to social capital formation. Finally, there is a need to repeat this study when new U.S. Census data become available.



## Endnotes

1. Including the effects of relationships in traditional economic models may be considered the domain of an emerging subdiscipline of economics and sociology referred to as socio-economics (Swedberg, 1990).
2. While James Coleman popularized the term social capital, Portes and Landolt credit Pierre Bourdieu with originating the term in the 1970's.
3. The references to social capital work cited here are not intended to discount the large amount of research that supports social capital conclusions but does not use the words "social capital." For example, important studies have focused on the importance of networks in reducing income differences between men and woman (Bartlett and Miller). This work might easily be interpreted as supporting the conclusion that social capital matters in determining salary levels.
4. We are grateful to Allan Schmid for his insights on the nature of externalities.
5. To model a high exclusion cost good, assume agents  $i$  and  $j$  begin the period with resources  $\bar{\alpha}$  and  $\bar{\beta}$ , respectively. Then, assume agents  $i$  and  $j$  invest  $\alpha$  and  $\beta$ , respectively, in a high exclusion cost good whose production of services is represented by the function  $f_h(\alpha+\beta)$ . Then, investments in the high exclusion cost good leave agents  $i$  and  $j$  with investments in individual goods of  $(\bar{\alpha} - \alpha) > 0$  and  $(\bar{\beta} - \beta) > 0$ , respectively. Income for agent  $i$  can be expressed as:
 
$$\pi_i(\alpha) = \pi_i[f_h(\alpha, \beta), (\bar{\alpha} - \alpha)]$$
6. To model a joint production model, assume agents  $i$  and  $j$  begin the period with resources  $\bar{\alpha}$  and  $\bar{\beta}$ , respectively.

Ownership of resources  $\bar{\alpha}$  and  $\bar{\beta}$  imply the right to create externalities. Also assume agents  $i$  and  $j$  invest  $\alpha$  and  $\beta$ , respectively, in the joint enterprise whose output is represented by the function  $f_j(\alpha, \beta)$  that is assumed to be increasing and concave down in  $\alpha$  and  $\beta$ . Then, assume investments in the joint enterprise leave agents  $i$  and  $j$  with investments in individual production activities of  $(\bar{\alpha} - \alpha)$  and  $(\bar{\beta} - \beta)$ , respectively. Assume agent  $i$  receives  $\gamma$  percent of the joint output. Income for agent  $i$  can be expressed as:

$$\pi_i(\alpha) = \pi_i[\gamma f_j(\alpha, \beta), \alpha]$$

7. To model a good owned in common, assume agents  $i$  and  $j$  initially extract services of  $\alpha$  and  $\beta$ , respectively, from a common resource. Also assume the cost of resource extraction is represented by  $f_c(\alpha, \beta)$  that is assumed to be an increasing and concave-down function of  $\alpha$  and  $\beta$ . Income for agent  $i$  can be expressed as:

$$\pi_i(\alpha) = \pi_i[f_c(\alpha, \beta), \alpha]$$

8. To model a ubiquitous externality model, assume agent  $i$  engages in individual production that not only earns an income of  $\pi_i(\alpha)$  but also diverts an input desired for alternative use by agent  $j$ . The negative value for agent  $j$  of the diverted input is  $f_j(\alpha)$ . Assume agent  $i$  chooses level of inputs  $\alpha$ . Then  $j$ 's income function can be written as:

$$\pi_j(\alpha, \beta) = \pi_j[\beta, f_j(\alpha)]$$

9. The conclusion that externally imposed income transfers will be offset to some degree by individual production decisions with external consequences is strengthened if the externally imposed income transfers reduce social capital.



For example, those forced to contribute to the welfare of a particular group may come to dislike the group which reduces the likelihood of voluntary efforts to redistribute income.

It is important to note the limitation of the conclusions reached in Appendix B. It has not been demonstrated that these results can be applied to an  $n$  person economy. It may be that persons are willing to contribute to the well-being of others if they know that others are contributing as well. Thus, an externally imposed tax might have the effect of encouraging more general support for income transfers. The main point here is that social capital needs to be included in any effort to examine the likely consequences of income transfer policies.

10. This approach was implemented by Pope, Kramer, Green, and Gardner to demonstrate the lack of robustness in land valuation models.
11. Data representing the percentage of households headed by a single parent with children were collected using the 1990 Census of Population and Housing U.S. Summary tape file 1C, U.S. Department of Commerce, Bureau of the Census, issued February 1992, CD90-1C.
12. For this purpose we used PUMS 1990. Household incomes are reported for 1989.
13. Factor analysis is a multivariate statistical method. Factor analysis is used to identify measures of underlying variables by analyzing the variation and cross-correlation within an observed variable set. This is accomplished through the generation of variables or factors that are highly correlated with some subset of the variables of interest and are independent of one another. Thus, factor analysis reduces an original set of indicator variables to a smaller set of underlying variables. The use of factor

analysis in this study is to reduce a comparatively large set of indicator variables to a few variables which can be used to predict means of household income and CVs.

14. Alternatively, the mean-standard deviation (mean-CV) efficient set is supported if one assumes perfect social capital and the distributions are related to each other by location-scale.
15. The other important point to notice about the results by race/ethnicity is how low are the CVs compared to the CVs reported by states. This is possible because part of the significant difference in the CV is accounted for by variations between the races/ethnicities. However, this does not explain the low total population CV. These differences may be explained in part by the use of different samples such as PUMS 1980, PUMS 1990, and CPS 1990/1991.

## Appendix A

### Income Distributions and Increases in Social Capital

To model the influence of relationships on economic activities, the potential influence  $j$  has on the terms of trade offered by  $i$  is represented by the social capital function  $k_{ij}(d_{ij}, r_{ij})$  where  $d_{ij}$  is the social distance and  $r_{ij}$  is the relationship (sympathy or antipathy) between agents  $i$  and  $j$ . If  $d_{ij}$  and  $r_{ij}$  are considered to be exogenous, then the relationship function is represented by the social capital coefficient  $k_{ij}$ . Otherwise, the social capital function  $k_{ij}(d_{ij}, r_{ij})$  is assumed to decrease with increases in  $d_{ij}$  for  $r_{ij} > 0$  and to increase with increases in  $r_{ij}$  for finite  $d_{ij}$  values. Positive values of  $k_{ij}$  reflect a resource for person  $j$  because an increase in  $j$ 's well-being increases  $i$ 's sense of well-being. So, an increase in  $k_{ij}$  increases person  $i$ 's willingness to offer  $j$  more favorable terms of trade, other things being equal.

Assume an economy consisting of two economic agents  $i$  and  $j$  whose preferences are described by ordinal utility functions  $U_i[\pi_i(\alpha), \pi_j(\alpha), k_{ij}]$ , and  $U_j[\pi_j(\alpha), \pi_i(\alpha), k_{ji}]$ , respectively. The arguments of the utility functions  $\pi_i(\alpha)$  and  $\pi_j(\alpha)$  represent income received by agents  $i$  and  $j$ , respectively, while  $k_{ij}$  and  $k_{ji}$  represent social capital coefficients. Furthermore, it is assumed that agent  $i$ 's income function is increasing and concave in his/her choice variable  $\alpha$  that has external consequences on agent  $j$ 's income function  $\pi_j(\alpha)$ . Finally, we assume that for  $0 < k_{ij} < 1$ , the following are true:

$$\frac{\partial^2 U_i}{\partial \pi_j \partial k_{ij}} > 0 \quad (\text{A-1a})$$

$$\frac{\partial^2 U_i}{\partial \pi_i \partial k_{ij}} = 0 \quad (\text{A-1b})$$

$$\frac{\partial^2 U_i}{\partial \pi_j \partial \pi_i} > 0 \quad (\text{A-1c})$$

The assumptions in the equations above imply the following for  $0 < k_{ij} < 1$ : (A-1a) implies that agent  $i$ 's marginal utility associated with an increase in agent  $j$ 's income increases with an increase in agent  $j$ 's social capital; (A-1b) implies that the marginal utility of own consumption is unaffected by an increase in agent  $j$ 's social capital; (A-1c) implies that incomes of agents  $i$  and  $j$  are complements in preferences.

Next, consider the economic consequences on agents  $i$  and  $j$ 's income and difference in incomes as social capital represented by social capital coefficient is increased. To begin, we write the first-order condition for agent  $i$  choosing his/her utility maximizing level of  $\alpha$  as:

$$\frac{dU_i}{d\alpha} = \frac{\partial U_i}{\partial \pi_i(\alpha)} \frac{\partial \pi_i(\alpha)}{\partial \alpha} + \frac{\partial U_i}{\partial \pi_j(\alpha)} \frac{\partial \pi_j(\alpha)}{\partial \alpha} = [\cdot] = 0 \quad (\text{A-2})$$



It is assumed that the second-order condition for  $\alpha$  holds so that  $\frac{\partial[\cdot]}{\partial \alpha} < 0$  and differentiating the first-order condition with respect to  $k_{ij}$  results in the expression:

$$\frac{d\alpha}{dk_{ij}} = \frac{-\frac{\partial^2 U_i}{\partial \pi_j(\alpha) \partial k_{ij}} \frac{\partial \pi_j(\alpha)}{\partial \alpha}}{\frac{\partial[\cdot]}{\partial \alpha}} \quad (\text{A-3})$$

From our earlier assumption, it follows that the sign of  $\frac{d\alpha}{dk_{ij}}$  depends on whether an increase in  $\alpha$  produces positive or negative externalities. If an increase in  $\alpha$  produces positive (negative) externalities, an increase in  $k_{ij}$  increases (decreases)  $\alpha$ .

Next, consider how an increase in social capital affects the total income of agents  $i$  and  $j$  equal to  $\pi_T = \pi_i(\alpha) + \pi_j(\alpha)$  and the difference in incomes equal to  $\pi_D = \pi_i(\alpha) - \pi_j(\alpha)$ . Maximizing  $\pi_T$  with respect to  $\alpha$  produces the result:

$$\frac{d\pi_T}{d\alpha} = \frac{\partial \pi_i(\alpha)}{\partial \alpha} + \frac{\partial \pi_j(\alpha)}{\partial \alpha} = 0 \quad (\text{A-4})$$

The value for  $\alpha$  that satisfies the first-order condition described above would never be chosen by agent  $i$  unless agent  $j$ 's social capital with agent  $i$  were sufficiently positive ( $k_{ij} > 0$ ). In fact, if agent  $j$  had no social capital with agent  $i$ , agent  $i$  would maximize his/her own income without regard to the externalities created by his/her choice of  $\alpha$ . Agent  $i$ 's selfishness of preference choice of  $\alpha$  would instead of satisfying the total income maximizing requirement would satisfy the requirement that  $\frac{\partial \pi_i(\alpha)}{\partial \alpha} = 0$ . But this choice, of course, would fail to maximize total income.

Assume that agent  $i$  has chosen his/her utility maximizing level of  $\alpha$ , namely  $\alpha^*$ . Next, consider the effect on  $\alpha^*$  of an increase in  $k_{ij}$ . As agent  $j$ 's social capital increases, the effect on total income can be expressed as:

$$\frac{d\pi_T}{dk_{ij}} = \left[ \frac{\partial \pi_i(\alpha^*)}{\partial \alpha^*} + \frac{\partial \pi_j(\alpha^*)}{\partial \alpha^*} \right] \frac{\partial \alpha}{\partial k_{ij}} \quad (\text{A-5})$$

Assume positive externalities and  $k_{ij} > 0$ ; then  $\frac{\partial \alpha}{\partial k_{ij}} = 0$  in equation (A-5). Next, consider the sign of the bracketed expression in equation (A-5) by comparing it with the expression in equation (A-2). If agent  $i$ 's marginal utility for own and agent  $j$ 's income were equal, then the bracketed expression in equation (A-5) must equal zero. On the other hand, if agent  $i$ 's marginal utility for own income were greater than (less than)

his/her marginal utility for agent  $j$ 's income, then the bracketed expression in equation (A-5) is positive (negative) and  $\alpha$  increases (decreases) with an increase in  $k_{ij}$ . It is generally accepted that agent  $i$  values a unit increase in his/her own income more than the same units of income increase for agent  $j$  unless there are wide differences in their relative incomes. Thus, under most conditions, the sign of equation (A-5) is unambiguously positive.

Next, consider the effect of an increase in social capital on difference in income  $\pi_D$ . If the relative income levels before the increase in  $k_{ij}$  are  $\pi_i(\alpha) > \pi_j(\alpha)$ , and an increase in  $k_{ij}$  lowers  $\pi_i(\alpha)$  and increases  $\pi_j(\alpha)$ , it follows that an increase in  $k_{ij}$  reduces the difference in incomes between agents  $i$  and  $j$ .

Having established the results above, an important income distribution conclusion can be deduced from our model:

*If  $\pi_i(\alpha) > \pi_j(\alpha)$  and  $0 < k_{ij} < 1$ , then increases in  $k_{ij}$  will increase total income and reduce the difference in incomes.*

If relative income levels before the increase in  $k_{ij}$  are  $\pi_i(\alpha) > \pi_j(\alpha)$ , and an increase in  $\alpha$  produces a much larger positive effect on  $\pi_j(\alpha)$  compared to a small reduction in  $\pi_i(\alpha)$ , then an increase in  $k_{ij}$  would increase the difference in incomes. This consequence leads to a second income distribution conclusion:

*If  $\pi_i(\alpha) < \pi_j(\alpha)$  and  $0 < k_{ij} < 1$ , then increases in  $k_{ij}$  will increase total income and reduce the difference in incomes.*



## Appendix B

### Income Transfers and Social Capital

One limitation of the social capital models described earlier is that the only means of redistributing income is through the production process. Indeed, in many business arrangements, this characterization of income redistribution possibilities may be accurate. However, in most advanced economic arrangements, there exist income redistribution possibilities in addition to production arrangements. One means of redistribution is transfer payments. Voluntary transfers have different effects on agent  $i$ 's choice of  $\alpha$  than do involuntary transfers.

Assume a transfer of income between agents  $i$  and  $j$  is required by the government to narrow the difference in their incomes. One might assume that if  $\pi_i(\alpha) > \pi_j(\alpha)$ , then the government might require agent  $i$  to transfer to agent  $j$  income of amount  $\delta$ .

Facing the possibility of a forced income transfer, agent  $i$ 's utility function is  $U_i[\pi_i(\alpha) - \delta, \pi_j(\alpha) + \delta, k_{ij}]$  and the first-order condition for  $i$ 's choice of  $\alpha$  is again that described in equation (A-3). Now consider agent  $i$ 's response to an increase in the required transfer  $\delta$ . Differentiating equation (A-3) with respect to  $\delta$ , we obtain:

$$\frac{d\alpha}{d\delta} = \frac{\frac{\partial^2 U_i}{\partial [\pi_i(\alpha) - \delta]^2} \frac{\partial \pi_i(\alpha)}{\partial \alpha} - \frac{\partial^2 U_j}{\partial [\pi_j(\alpha) - \delta]^2} \frac{\partial \pi_j(\alpha)}{\partial \alpha} - \frac{\partial^2 U_i}{\partial [\pi_j(\alpha) + \delta] \partial [\pi_i - \delta]} \left[ \frac{\partial \pi_i(\alpha)}{\partial \alpha} - \frac{\partial \pi_j(\alpha)}{\partial \alpha} \right]}{\frac{\partial [\cdot]}{\partial \alpha}} \quad (\text{B-1a})$$

for  $\delta > 0$  where:

$$\frac{\partial \pi_i(\alpha)}{\partial \alpha} \leq 0, \left[ \frac{\partial \pi_i(\alpha)}{\partial \alpha} - \frac{\partial \pi_j(\alpha)}{\partial \alpha} \right] < 0, \text{ and } \frac{d\alpha}{d\delta} < 0. \quad (\text{B-1b})$$

If increasing  $\alpha$  produces positive externalities for agent  $j$ , then

$$\frac{\partial \pi_j(\alpha)}{\partial \alpha} > 0, \frac{\partial \pi_i(\alpha)}{\partial \delta} = -1, \frac{\partial [\pi_j(\alpha) + \delta]}{\partial \delta} = 1, \text{ and } \frac{d\alpha}{d\delta} < 0. \quad (\text{B-1c})$$

Equation B-1a suggests that redistribution of income through externally imposed transfers reduces the transfers previously made by increases in  $\alpha$  that increase total income and reduce differences in income.

If increasing  $\alpha$  produces negative externalities for agent  $j$ , then  $\frac{d\alpha}{d\delta} > 0$ .

On the other hand, if  $\delta < 0$ , requiring a transfer from agent  $j$  to agent  $i$ , then  $\frac{d\alpha}{d\delta} > 0$  for positive externalities and  $\frac{d\alpha}{d\delta} < 0$  for negative externalities. Again, the implications are that externally imposed redistributive efforts will be offset by redistribution efforts made through production decisions. Furthermore, the results would be strengthened if the externally imposed transfers reduced social capital.

In contrast to required redistributions, suppose that the redistributive efforts were voluntary and agent  $i$  chooses  $\delta$  to maximize his/her own utility. Provided that  $\pi_i(\alpha) > \pi_j(\alpha)$ , the assumptions in equation (A-1) imply that:

$$\frac{d\delta}{dk_{ij}} > 0 \quad (B-2)$$

suggesting that increases in social capital will unambiguously reduce the difference in incomes between agents  $i$  and  $j$  when  $\pi_i(\alpha) > \pi_j(\alpha)$ .



# **Appendix C** **Social Capital Indicator Variables and Income** **Distributions for the U.S., 1990**

**Table C1. Correlation Coefficients Between Indicator Variables and Coefficients of Variation by State, 1990**

| Variables   | Correlation Coefficients | Significance Levels |
|---|--------------------------|---------------------|
| Family (F90)  |                          |                     |
| Percentages of Households Headed by a Single Female with Children (F90/HHSFC) | .4013                    | .004                |
| Birth Rates of Single Teens (F90/BRST)  | .5396                    | .000                |
| Infant Mortality Rates (F90/IMR)  | .3038                    | .032                |
| Education (E90)   |                          |                     |
| High School Graduation Rates (E90/HSGR)                                       | -.3413                   | .015                |
| Percentage of Teens Not in School (E90/TNIS)                                  | .3867                    | .006                |
| Crime (C90)   |                          |                     |
| Litigation Rates (C90/LIT)  | .4985                    | .000                |
| Violent Death Rates for Teens (C90/VDT)                                       | .5228                    | .000                |
| Labor (L90)   |                          |                     |
| Labor Force Participation Rates (L90/LFPR)                                    | -.6800                   | .000                |
| Child Poverty Rates (L90/CPR)   | .7572                    | .000                |

Source: Estimated by the authors.

**Table C2. Correlation Coefficients Between Indicator Variables and Means of Household Income by State, 1990**

| Variables   | Correlation Coefficients | Significance Levels |
|---|--------------------------|---------------------|
| Family (F90)  |                          |                     |
| Percentages of Households Headed by a Single Female with Children (F90/HHSFC) | .0268                    | .853                |
| Birth Rates of Single Teens (F90/BRST)  | -.4135                   | .003                |
| Infant Mortality Rates (F90/IMR)  | -.1323                   | .360                |
| Education (E90)   |                          |                     |
| High School Graduation Rates (E90/HSGR)                                       | -.0365                   | .801                |
| Percentage of Teens Not in School (E90/TNIS)                                  | -.1053                   | .467                |
| Crime (C90)   |                          |                     |
| Litigation Rates (C90/LIT)  | -.2203                   | .124                |
| Violent Death Rates for Teens (C90/VDT)                                       | -.4436                   | .001                |
| Labor (L90)   |                          |                     |
| Labor Force Participation Rates (L90/LFPR)                                    | .5764                    | .000                |
| Child Poverty Rates (L90/CPR)   | -.4620                   | .001                |

Source: Estimated by the authors.

**Table C3. Rotated Factor Matrix of Indicator Variables for States, 1990**

| Variables       | Factor 1<br>(Education) | Factor 2<br>(Family) | Factor 3<br>(Labor) | Factor 4<br>(Crime) |
|-----------------|-------------------------|----------------------|---------------------|---------------------|
| Education (E90) |                         |                      |                     |                     |
| E90/HSGR        | <b>-.8561</b>           | -.3236               | -.0971              | -.0708              |
| E90/TNIS        | <b>.7646</b>            | .0268                | .2648               | .4149               |
| Family (F90)    |                         |                      |                     |                     |
| F90/IMR         | .1318                   | <b>.8894</b>         | .1070               | .2918               |
| F90/BRST        | .2758                   | <b>.6594</b>         | .4795               | .3492               |
| F90/HHSFC       | .5289                   | <b>.6268</b>         | .3161               | .0086               |
| Labor (L90)     |                         |                      |                     |                     |
| L90/LFPR        | -.0792                  | -.1250               | <b>-.9524</b>       | -.1268              |
| L90/CPR         | .4296                   | .2957                | <b>.6838</b>        | .3564               |
| Crime (C90)     |                         |                      |                     |                     |
| C90/VDT         | .0712                   | .1383                | .1404               | <b>.8994</b>        |
| C90/LIT         | .2407                   | .3283                | .2143               | <b>.6049</b>        |

Cumulation percentage of variance for the four factors equals 83.5.

Source: Estimated by the authors.

**Table C4. Regression Analysis to Predict Coefficients of Variation for States, 1990**

| Independent Variables   | Betas        | T-Statistics | Significance Levels |
|-------------------------|--------------|--------------|---------------------|
| Factor 1 (Education)    | .013685      | 2.410        | .0202               |
| Factor 2 (Family)       | .009545      | 1.712        | .0939               |
| Factor 3 (Labor )       | .036230      | 6.210        | .0000               |
| Factor 4 (Crime)        | .024581      | 4.398        | .0001               |
| T90/H (Health Transfer) | -2.04930E-04 | -1.959       | .0564               |
| Constant                | 1.034320     | 55.672       | .0000               |
| Adj R <sup>2</sup>      | .61948       |              |                     |
| F-Statistic             | 16.95448     |              | .0000               |

Source: Estimated by the authors.



**Table C5. Regression Analysis to Predict Means of Household Income for States, 1990**

| <b>Independent Variables</b> | <b>Betas</b> | <b>T-Statistics</b> | <b>Significance Levels</b> |
|------------------------------|--------------|---------------------|----------------------------|
| Factor 1 (Education)         | 953.503      | 1.661               | .104                       |
| Factor 2 (Family)            | -147.225     | -.261               | .795                       |
| Factor 3 (Labor)             | -2949.375    | -5.000              | .000                       |
| Factor 4 (Crime)             | -2043.523    | -3.616              | .001                       |
| T90/H (Health Transfer)      | 29.929       | 2.830               | .007                       |
| (Constant)                   | 26123.647    | 13.907              | .000                       |
| <hr/>                        |              |                     |                            |
| Adj R <sup>2</sup>           | 0.568        |                     |                            |
| F-Statistic                  | 13.883       |                     | .000                       |

Source: Estimated by the authors.

**Table C6. Coefficients of Variation in Descending Order and Means of Household Income by State, 1990**

| States         | Coefficients of Variation | Means (\$) | Standard Deviations (\$) |
|----------------|---------------------------|------------|--------------------------|
| Louisiana      | 1.14680                   | 24,685.20  | 28,309.10                |
| Florida        | 1.11346                   | 29,198.36  | 32,511.08                |
| Arizona        | 1.10437                   | 28,095.19  | 31,027.36                |
| Oklahoma       | 1.09976                   | 25,295.50  | 27,818.93                |
| Mississippi    | 1.09936                   | 23,494.57  | 25,829.09                |
| Texas          | 1.09316                   | 29,544.71  | 32,297.14                |
| Arkansas       | 1.08324                   | 23,662.03  | 25,631.54                |
| West Virginia  | 1.07543                   | 23,083.16  | 24,824.39                |
| New Mexico     | 1.06145                   | 25,626.48  | 27,201.14                |
| Alabama        | 1.05617                   | 26,660.69  | 28,158.15                |
| Maine          | 1.05337                   | 25,934.26  | 27,318.37                |
| New York       | 1.04383                   | 38,640.20  | 40,333.75                |
| Idaho          | 1.04358                   | 26,744.90  | 27,910.51                |
| Kentucky       | 1.03860                   | 26,216.66  | 27,228.68                |
| Tennessee      | 1.03669                   | 28,296.34  | 29,334.50                |
| Vermont        | 1.03311                   | 26,628.33  | 27,510.08                |
| Georgia        | 1.02783                   | 31,948.28  | 32,837.38                |
| Montana        | 1.02708                   | 23,492.22  | 24,128.42                |
| North Dakota   | 1.02080                   | 24,639.44  | 25,152.01                |
| Missouri       | 1.02066                   | 28,913.93  | 29,511.36                |
| South Carolina | 1.00911                   | 27,698.94  | 27,951.41                |
| North Carolina | 1.00560                   | 28,886.45  | 29,048.35                |
| Pennsylvania   | 1.00208                   | 32,242.97  | 32,310.05                |
| Wyoming        | 1.00144                   | 26,176.89  | 26,214.66                |
| Colorado       | 0.99680                   | 31,441.86  | 31,341.28                |
| Kansas         | 0.99353                   | 30,070.70  | 29,876.24                |
| Nevada         | 0.98639                   | 33,406.04  | 32,951.49                |
| Michigan       | 0.98505                   | 33,010.67  | 32,517.16                |
| Nebraska       | 0.97294                   | 28,384.98  | 27,616.83                |



| States        | Coefficients of Variation | Means (\$) | Standard Deviations (\$) |
|---------------|---------------------------|------------|--------------------------|
| Illinois      | 0.96993                   | 36,514.68  | 35,416.72                |
| South Dakota  | 0.96767                   | 23,953.19  | 23,178.68                |
| Delaware      | 0.96662                   | 34,528.36  | 33,375.85                |
| Minnesota     | 0.96472                   | 32,736.85  | 31,582.05                |
| California    | 0.95898                   | 40,958.99  | 39,279.01                |
| Alaska        | 0.94956                   | 39,065.86  | 37,095.44                |
| New Jersey    | 0.94953                   | 44,566.43  | 42,317.13                |
| Ohio          | 0.94894                   | 32,078.45  | 30,440.46                |
| New Hampshire | 0.94716                   | 33,709.38  | 31,928.28                |
| Wisconsin     | 0.94710                   | 30,616.95  | 28,997.44                |
| Massachusetts | 0.94637                   | 39,959.38  | 37,816.55                |
| Oregon        | 0.94603                   | 30,699.03  | 29,042.18                |
| Rhode Island  | 0.94458                   | 34,682.38  | 32,760.31                |
| Connecticut   | 0.93719                   | 46,950.29  | 44,001.12                |
| Iowa          | 0.93219                   | 29,137.47  | 27,161.71                |
| Indiana       | 0.92986                   | 31,214.10  | 29,024.89                |
| Virginia      | 0.92821                   | 37,047.50  | 34,388.01                |
| Washington    | 0.91118                   | 33,937.52  | 30,923.10                |
| Utah          | 0.90398                   | 30,614.17  | 27,674.74                |
| Hawaii        | 0.89995                   | 42,325.21  | 38,090.69                |
| Maryland      | 0.89492                   | 42,821.59  | 38,322.11                |

Source: Data used to calculate values in other columns were obtained from U.S. Department of Commerce, Bureau of the Census, 1990 Census of Population and Housing, Public Use Microdata Sample (PUMS), CD90AA, August 1993.

**Table C7. Means of Household Income in Descending Order and Coefficients of Variation by State, 1990**

| States        | Means (\$) | Standard Deviations (\$) | Coefficients of Variation |
|---------------|------------|--------------------------|---------------------------|
| Connecticut   | 46,950.29  | 44,001.12                | 0.93719                   |
| New Jersey    | 44,566.43  | 42,317.13                | 0.94953                   |
| Maryland      | 42,821.59  | 38,322.11                | 0.89492                   |
| Hawaii        | 42,325.21  | 38,090.69                | 0.89995                   |
| California    | 40,958.99  | 39,279.01                | 0.95898                   |
| Massachusetts | 39,959.38  | 37,816.55                | 0.94637                   |
| Alaska        | 39,065.86  | 37,095.44                | 0.94956                   |
| New York      | 38,640.20  | 40,333.75                | 1.04383                   |
| Virginia      | 37,047.50  | 34,388.01                | 0.92821                   |
| Illinois      | 36,514.68  | 35,416.72                | 0.96993                   |
| Rhode Island  | 34,682.38  | 32,760.31                | 0.94458                   |
| Delaware      | 34,528.36  | 33,375.85                | 0.96662                   |
| Washington    | 33,937.52  | 30,923.10                | 0.91118                   |
| New Hampshire | 33,709.38  | 31,928.28                | 0.94716                   |
| Nevada        | 33,406.04  | 32,951.49                | 0.98639                   |
| Michigan      | 33,010.67  | 32,517.16                | 0.98505                   |
| Minnesota     | 32,736.85  | 31,582.05                | 0.96472                   |
| Pennsylvania  | 32,242.97  | 32,310.05                | 1.00208                   |
| Ohio          | 32,078.45  | 30,440.46                | 0.94894                   |
| Georgia       | 31,948.28  | 32,837.38                | 1.02783                   |
| Colorado      | 31,441.86  | 31,341.28                | 0.99680                   |
| Indiana       | 31,214.10  | 29,024.89                | 0.92986                   |
| Oregon        | 30,699.03  | 29,042.18                | 0.94603                   |
| Wisconsin     | 30,616.95  | 28,997.44                | 0.94710                   |
| Utah          | 30,514.17  | 27,674.74                | 0.90398                   |
| Kansas        | 30,070.70  | 29,876.24                | 0.99353                   |
| Texas         | 29,544.71  | 32,297.14                | 1.09316                   |
| Florida       | 29,198.36  | 32,511.08                | 1.11346                   |
| Iowa          | 29,137.47  | 27,161.71                | 0.93219                   |
| Missouri      | 28,913.93  | 29,511.36                | 1.02066                   |



| States         | Means (\$) | Standard Deviations (\$) | Coefficients of Variation |
|----------------|------------|--------------------------|---------------------------|
| North Carolina | 28,886.45  | 29,048.35                | 1.00560                   |
| Nebraska       | 28,384.98  | 27,616.83                | 0.97294                   |
| Tennessee      | 28,296.34  | 29,334.55                | 1.03669                   |
| Arizona        | 28,095.19  | 31,027.36                | 1.10437                   |
| South Carolina | 27,698.94  | 27,951.41                | 1.00911                   |
| Idaho          | 26,744.90  | 27,910.51                | 1.04358                   |
| Alabama        | 26,660.69  | 28,158.15                | 1.05617                   |
| Vermont        | 26,628.33  | 27,510.08                | 1.03311                   |
| Kentucky       | 26,216.66  | 27,228.68                | 1.03860                   |
| Wyoming        | 26,176.89  | 26,214.66                | 1.00144                   |
| Maine          | 25,934.26  | 27,318.37                | 1.05337                   |
| New Mexico     | 25,626.48  | 27,201.14                | 1.06145                   |
| Oklahoma       | 25,295.50  | 27,818.93                | 1.09976                   |
| Louisiana      | 24,685.20  | 28,309.10                | 1.14680                   |
| North Dakota   | 24,639.44  | 25,152.01                | 1.02080                   |
| South Dakota   | 23,953.19  | 23,178.68                | 0.96767                   |
| Arkansas       | 23,662.03  | 25,631.54                | 1.08324                   |
| Mississippi    | 23,494.57  | 25,829.09                | 1.09936                   |
| Montana        | 23,492.22  | 24,128.42                | 1.02708                   |
| West Virginia  | 23,083.16  | 24,824.39                | 1.07543                   |

Source: Same as Table C6.

**Table C8. Regression Analysis: Means of Household Income on Coefficients of Variation by State, 1990**

| Independent Variables    | Betas      | T-Statistics | Significance Levels |
|--------------------------|------------|--------------|---------------------|
| Constant                 | 95602.416  | 7.803        | .000                |
| Coefficient of Variation | -65616.982 | -5.363       | .000                |
| <hr/>                    |            |              |                     |
| Adjusted R <sup>2</sup>  | 0.362      |              |                     |
| F-Statistic              | 28.765     |              | .000                |

Source: Estimated by the authors.

# **Appendix D** **Social Capital Indicator Variables and Income** **Distributions for the U.S., 1980**

**Table D1. Correlation Coefficients Between Indicator Variables and Coefficients of Variation by State, 1980**

| Variables   | Correlation Coefficients | Significance Levels |
|---|--------------------------|---------------------|
| Family (F80)  |                          |                     |
| Percentages of Households Headed by a Single Female with Children (F80/HHSFC) | -.0441                   | .761                |
| Birth Rates of Single Teens (F80/BRST)  | .7008                    | .000                |
| Infant Mortality Rates (F80/IMR)  | .5480                    | .000                |
| Education (E80)   |                          |                     |
| High School Graduation Rates (E80/HSGR)                                       | -.7245                   | .000                |
| Percentage of Teens Not in School (E80/TNIS)                                  | .5055                    | .000                |
| Crime (C80)   |                          |                     |
| Litigation Rates (C80/LIT)  | .3884                    | .005                |
| Violent Death Rates for Teens (C80/VDT)                                       | .1750                    | .224                |
| Labor (L80)   |                          |                     |
| Labor Force Participation Rates (L80/LFPR)                                    | -.6716                   | .000                |
| Child Poverty Rates (L80/CPR)   | .8579                    | .000                |

Source: Estimated by the authors.

**Table D2. Correlation Coefficients Between Indicator Variables and Means of Household Income by State, 1980**

| Variables   | Correlation Coefficients | Significance Levels |
|---|--------------------------|---------------------|
| Family (F80)  |                          |                     |
| Percentages of Households Headed by a Single Female with Children (F80/HHSFC) | .5454                    | .000                |
| Birth Rates of Single Teens (F80/BRST)  | -.4948                   | .000                |
| Infant Mortality Rates (F80/IMR)  | -.1984                   | .167                |
| Education (E80)   |                          |                     |
| High School Graduation Rates (E80/HSGR)                                       | .5948                    | .000                |
| Percentage of Teens Not in School E80/TNIS)                                   | -.2782                   | .050                |
| Crime (C80)   |                          |                     |
| Litigation Rates (C80/LIT)  | .0120                    | .934                |
| Violent Death Rates for Teens (C80/VDT)                                       | .0356                    | .806                |
| Labor (L80)   |                          |                     |
| Labor Force Participation Rates (L80/LFPR)                                    | .6871                    | .000                |
| Child Poverty Rates (L80/CPR)   | -.5798                   | .000                |

Source: Estimated by the authors.



**Table D3. Rotated Factor Matrix of Indicator Variables for States, 1980**

| <b>Variables</b>             | <b>Factor 1<br/>(Labor/Educ.)</b> | <b>Factor 2<br/>(Educ./Family)</b> | <b>Factor 3<br/>(Family)</b> | <b>Factor 4<br/>(Crime)</b> |
|------------------------------|-----------------------------------|------------------------------------|------------------------------|-----------------------------|
| Labor (L80)/Education (E80)  |                                   |                                    |                              |                             |
| LFPR                         | <b>-.87708</b>                    | -.23302                            | .19077                       | -.01254                     |
| HSGR                         | <b>-.67739</b>                    | -.57950                            | -.20207                      | .22666                      |
| CPR                          | <b>.67622</b>                     | .52716                             | .28726                       | .13313                      |
| Education (E80)/Family (F80) |                                   |                                    |                              |                             |
| TNIS                         | .10869                            | <b>.84615</b>                      | .08316                       | .39923                      |
| BRST                         | .50492                            | <b>.76614</b>                      | .16445                       | .21639                      |
| Family (F80)                 |                                   |                                    |                              |                             |
| HHSFC                        | -.23038                           | .04698                             | <b>.89485</b>                | .17802                      |
| IMR                          | .48537                            | .44383                             | <b>.60159</b>                | -.06742                     |
| Crime (C80)                  |                                   |                                    |                              |                             |
| VDT                          | -.14178                           | .20633                             | .05573                       | <b>.90068</b>               |
| LIT                          | .41569                            | .01750                             | .53321                       | <b>.62952</b>               |

Cumulation percentage of variance for the four factors equals 86.8.

Source: Estimated by the authors.

**Table D4. Regression Analysis to Predict Coefficients of Variation for States, 1980**

| <b>Independent Variables</b>   | <b>Betas</b> | <b>T-Statistics</b> | <b>Significance Levels</b> |
|--------------------------------|--------------|---------------------|----------------------------|
| Factor 1<br>(Labor/Education)  | .030693      | 9.653               | .0000                      |
| Factor 2<br>(Education/Family) | .023265      | 6.755               | .0000                      |
| Factor 3 (Family)              | .001138      | .365                | .7167                      |
| Factor 4 (Crime)               | .008941      | 2.796               | .0076                      |
| T80/H (Health Transfer)        | 1.05960E-04  | 2.762               | .0083                      |
| Constant                       | .673225      | 24.999              | .0000                      |
| Adj R <sup>2</sup>             | .72856       |                     |                            |
| F-Statistic                    | 27.30409     |                     | .0000                      |

Source: Estimated by the authors.

**Table D5. Regression Analysis to Predict Means of Household Income by States, 1980**

| <b>Independent Variables</b>   | <b>Betas</b> | <b>T-Statistics</b> | <b>Significance Levels</b> |
|--------------------------------|--------------|---------------------|----------------------------|
| Factor 1<br>(Labor/Education)  | -1623.670    | -7.892              | .000                       |
| Factor 2<br>(Education/Family) | -821.382     | -3.685              | .001                       |
| Factor 3 (Family)              | 1100.546     | 5.456               | .000                       |
| Factor 4 (Crime)               | 162.263      | .784                | .437                       |
| T80/H (Health Transfer)        | 2.522        | 1.016               | .315                       |
| (Constant)                     | 17444.218    | 10.01               | .000                       |
| <hr/>                          |              |                     |                            |
| Adj R <sup>2</sup>             | .717         |                     |                            |
| F-Statistic                    | 25.820       |                     | .000                       |

Source: Estimated by the authors.



**Table D6. Coefficients of Variation in Descending Order and Means of Household Income by State, 1980**

| States         | Coefficients of Variation | Means (\$) | Standard Deviations (\$) |
|----------------|---------------------------|------------|--------------------------|
| Mississippi    | 0.84486                   | 15,051.39  | 12,716.31                |
| Arkansas       | 0.82561                   | 14,956.46  | 12,348.17                |
| Louisiana      | 0.81903                   | 17,958.07  | 14,708.27                |
| Alabama        | 0.81179                   | 16,599.88  | 13,475.54                |
| Florida        | 0.80842                   | 18,017.00  | 14,565.24                |
| Kentucky       | 0.80700                   | 16,638.07  | 13,426.96                |
| Tennessee      | 0.80041                   | 16,875.31  | 13,507.24                |
| Georgia        | 0.79775                   | 17,971.36  | 14,336.62                |
| Oklahoma       | 0.79232                   | 19,642.10  | 13,978.15                |
| South Dakota   | 0.78325                   | 15,793.57  | 12,370.37                |
| New York       | 0.78076                   | 19,780.19  | 15,443.63                |
| Rhode Island   | 0.77326                   | 17,565.20  | 13,582.46                |
| West Virginia  | 0.77097                   | 16,739.84  | 12,905.90                |
| South Carolina | 0.76979                   | 17,145.57  | 13,198.55                |
| Texas          | 0.76747                   | 19,772.89  | 15,175.11                |
| New Mexico     | 0.76537                   | 17,308.64  | 13,247.51                |
| North Dakota   | 0.76493                   | 16,071.68  | 12,293.67                |
| Nebraska       | 0.76451                   | 17,827.39  | 13,629.13                |
| Missouri       | 0.76348                   | 18,276.32  | 13,853.53                |
| Montana        | 0.76231                   | 17,486.44  | 13,330.01                |
| North Carolina | 0.76140                   | 16,753.05  | 12,755.80                |
| Kansas         | 0.74871                   | 18,991.08  | 14,218.76                |
| California     | 0.74733                   | 21,528.72  | 16,089.00                |
| Idaho          | 0.74623                   | 17,759.74  | 13,252.88                |
| Delaware       | 0.74364                   | 20,690.29  | 15,386.20                |
| Arizona        | 0.73869                   | 19,292.59  | 14,251.17                |
| Nevada         | 0.73330                   | 20,845.14  | 15,285.76                |
| Virginia       | 0.73274                   | 20,580.44  | 15,080.05                |
| Massachusetts  | 0.73130                   | 20,287.21  | 14,835.96                |
| Maine          | 0.73088                   | 15,958.13  | 11,663.40                |
| Iowa           | 0.72987                   | 19,009.19  | 13,874.19                |
| Minnesota      | 0.72928                   | 20,143.52  | 14,690.36                |

| States        | Coefficients of Variation | Means (\$) | Standard Deviations (\$) |
|---------------|---------------------------|------------|--------------------------|
| Oregon        | 0.72884                   | 19,320.03  | 14,081.20                |
| Pennsylvania  | 0.72827                   | 19,233.31  | 14,007.12                |
| Colorado      | 0.72463                   | 21,046.37  | 15,250.80                |
| Illinois      | 0.72157                   | 21,830.70  | 15,752.33                |
| Vermont       | 0.71851                   | 17,088.57  | 12,278.33                |
| Connecticut   | 0.71703                   | 23,056.26  | 16,532.07                |
| New Jersey    | 0.71467                   | 22,512.20  | 16,088.73                |
| Ohio          | 0.71102                   | 19,937.08  | 14,175.57                |
| Michigan      | 0.70921                   | 21,546.55  | 15,281.00                |
| Washington    | 0.70919                   | 20,863.66  | 14,796.23                |
| Hawaii        | 0.70330                   | 23,585.46  | 16,587.71                |
| Indiana       | 0.70261                   | 19,730.91  | 13,863.10                |
| Maryland      | 0.69889                   | 23,085.55  | 16,134.37                |
| Wisconsin     | 0.69340                   | 19,939.43  | 13,825.94                |
| New Hampshire | 0.68686                   | 19,242.23  | 13,216.74                |
| Utah          | 0.68301                   | 19,907.37  | 13,596.91                |
| Wyoming       | 0.68176                   | 21,619.59  | 14,739.34                |
| Alaska        | 0.67784                   | 29,304.99  | 19,864.09                |

Source: Same as Table C6, but for 1980.



**Table D7. Means of Household Income in Descending Order and Coefficients of Variation by State, 1980**

| States        | Means (\$) | Standard Deviations (\$) | Coefficients of Variation |
|---------------|------------|--------------------------|---------------------------|
| Alaska        | 29,304.99  | 19,864.09                | 0.67784                   |
| Hawaii        | 23,585.46  | 16,587.71                | 0.70330                   |
| Maryland      | 23,085.55  | 16,134.37                | 0.69889                   |
| Connecticut   | 23,056.26  | 16,532.07                | 0.71703                   |
| New Jersey    | 22,512.20  | 16,088.73                | 0.71467                   |
| Illinois      | 21,830.70  | 15,752.33                | 0.72157                   |
| Wyoming       | 21,619.59  | 14,739.34                | 0.68176                   |
| Michigan      | 21,546.55  | 15,281.00                | 0.70921                   |
| California    | 21,528.72  | 16,089.00                | 0.74733                   |
| Colorado      | 21,046.37  | 15,250.80                | 0.72463                   |
| Washington    | 20,863.66  | 14,796.23                | 0.70919                   |
| Nevada        | 20,845.14  | 15,285.76                | 0.73330                   |
| Delaware      | 20,690.29  | 15,386.20                | 0.74364                   |
| Virginia      | 20,580.44  | 15,080.05                | 0.73274                   |
| Massachusetts | 20,287.21  | 14,835.96                | 0.73130                   |
| Minnesota     | 20,143.52  | 14,690.36                | 0.72928                   |
| Wisconsin     | 19,939.43  | 13,825.94                | 0.69340                   |
| Ohio          | 19,937.08  | 14,175.57                | 0.71102                   |
| Utah          | 19,907.37  | 13,596.91                | 0.68301                   |
| New York      | 19,780.19  | 15,443.63                | 0.78076                   |
| Texas         | 19,772.89  | 15,175.11                | 0.76747                   |
| Indiana       | 19,730.91  | 13,863.10                | 0.70261                   |
| Oklahoma      | 19,642.10  | 13,978.15                | 0.79232                   |
| Oregon        | 19,320.03  | 14,081.20                | 0.72884                   |
| Arizona       | 19,292.59  | 14,251.17                | 0.73869                   |
| New Hampshire | 19,242.23  | 13,216.74                | 0.68686                   |
| Pennsylvania  | 19,233.31  | 14,007.12                | 0.72827                   |
| Iowa          | 19,009.19  | 13,874.19                | 0.72987                   |
| Kansas        | 18,991.08  | 14,218.76                | 0.74871                   |
| Missouri      | 18,276.32  | 13,953.53                | 0.76348                   |
| Florida       | 18,017.00  | 14,565.24                | 0.80842                   |
| Georgia       | 17,971.36  | 14,336.62                | 0.79775                   |
| Louisiana     | 17,958.07  | 14,708.27                | 0.81903                   |

| States         | Means (\$) | Standard Deviations (\$) | Coefficients of Variation |
|----------------|------------|--------------------------|---------------------------|
| Nebraska       | 17,827.39  | 13,629.13                | 0.76451                   |
| Idaho          | 17,759.74  | 13,252.88                | 0.74623                   |
| Rhode Island   | 17,565.20  | 13,582.46                | 0.77326                   |
| Montana        | 17,486.44  | 13,330.01                | 0.76231                   |
| New Mexico     | 17,308.64  | 13,247.51                | 0.76537                   |
| South Carolina | 17,145.57  | 13,198.55                | 0.76979                   |
| Vermont        | 17,088.57  | 12,278.33                | 0.71851                   |
| Tennessee      | 16,875.31  | 13,507.24                | 0.80041                   |
| North Carolina | 16,753.05  | 12,755.80                | 0.76140                   |
| West Virginia  | 16,739.84  | 12,905.90                | 0.77097                   |
| Kentucky       | 16,638.07  | 13,426.96                | 0.80700                   |
| Alabama        | 16,599.88  | 13,475.54                | 0.81179                   |
| North Dakota   | 16,071.68  | 12,293.67                | 0.76493                   |
| Maine          | 15,958.13  | 11,663.40                | 0.73088                   |
| South Dakota   | 15,793.57  | 12,370.37                | 0.78325                   |
| Mississippi    | 15,051.39  | 12,716.31                | 0.84486                   |
| Arkansas       | 14,956.46  | 12,348.17                | 0.82561                   |

Source: Same as Table C6, but for 1980.

**Table D8. Regression Analysis: Means of Household Income on Coefficients of Variation by State, 1980**

| Independent Variables    | Betas      | T-Statistics | Significance Levels |
|--------------------------|------------|--------------|---------------------|
| Constant                 | 43,917.71  | 10.942       | .000                |
| Coefficient of Variation | -34,026.80 | -6.343       | .000                |
| Adjusted R <sup>2</sup>  | .445       |              |                     |
| F-Statistic              | 40.238     |              | .000                |

Source: Estimated by the authors.



## Appendix E

### U.S. Income Distribution Changes Between 1980 and 1990

**Table E1. Increase in Means of Household Income and Coefficients of Variation (CV) by State in Descending Order, 1980 to 1990**

| States         | Mean Income Ratios<br>1990/1980 | CV Ratios<br>1990/1980 | CVs<br>1990 | CVs<br>1980 | Mean Incomes<br>(\$)<br>1990 | Mean Incomes<br>(\$)<br>1980 <sup>a</sup> |
|----------------|---------------------------------|------------------------|-------------|-------------|------------------------------|---|
| Connecticut    | 1.28071                         | 1.30704                | 0.93719     | 0.71703     | 46,950.29                    | 36,659.45                                 |
| New Jersey     | 1.24507                         | 1.32863                | 0.94953     | 0.71467     | 44,566.43                    | 35,794.40                                 |
| Rhode Island   | 1.24182                         | 1.22156                | 0.94458     | 0.77326     | 34,682.38                    | 27,928.67                                 |
| Massachusetts  | 1.23879                         | 1.29409                | 0.94637     | 0.73130     | 39,959.38                    | 32,256.66                                 |
| New York       | 1.22860                         | 1.33694                | 1.04383     | 0.78076     | 38,640.20                    | 31,450.50                                 |
| California     | 1.19656                         | 1.28321                | 0.95898     | 0.74733     | 40,958.99                    | 34,230.66                                 |
| Maryland       | 1.16661                         | 1.28049                | 0.89492     | 0.69889     | 42,821.59                    | 36,706.02                                 |
| Virginia       | 1.13216                         | 1.26677                | 0.92821     | 0.73274     | 37,047.50                    | 32,722.90                                 |
| Hawaii         | 1.12865                         | 1.27961                | 0.89995     | 0.70330     | 42,325.21                    | 37,500.88                                 |
| Georgia        | 1.11807                         | 1.28841                | 1.02783     | 0.79775     | 31,948.28                    | 28,574.46                                 |
| New Hampshire  | 1.10179                         | 1.37897                | 0.94716     | 0.68686     | 33,709.38                    | 30,595.15                                 |
| Alabama        | 1.10110                         | 1.30104                | 1.05617     | 0.81179     | 26,660.69                    | 26,393.81                                 |
| North Carolina | 1.08443                         | 1.32072                | 1.00560     | 0.76140     | 28,886.45                    | 26,637.35                                 |
| Tennessee      | 1.05458                         | 1.29520                | 1.03669     | 0.80041     | 28,296.34                    | 26,831.74                                 |
| Pennsylvania   | 1.05435                         | 1.37597                | 1.00208     | 0.72827     | 32,242.97                    | 30,580.96                                 |
| Illinois       | 1.05197                         | 1.34419                | 0.96993     | 0.72157     | 36,514.68                    | 34,710.81                                 |
| Delaware       | 1.04957                         | 1.29985                | 0.96662     | 0.74364     | 34,528.36                    | 32,897.56                                 |
| Washington     | 1.02304                         | 1.28482                | 0.91118     | 0.70919     | 33,937.52                    | 33,173.22                                 |
| Minnesota      | 1.02213                         | 1.32284                | 0.96472     | 0.72928     | 32,736.85                    | 32,028.20                                 |
| Maine          | 1.02210                         | 1.44124                | 1.05337     | 0.73088     | 25,934.26                    | 25,373.43                                 |
| Florida        | 1.01925                         | 1.37733                | 1.11346     | 0.80842     | 29,198.36                    | 28,647.03                                 |
| South Carolina | 1.01605                         | 1.31089                | 1.00911     | 0.76979     | 27,698.94                    | 27,261.46                                 |
| Ohio           | 1.01194                         | 1.33462                | 0.94894     | 0.71102     | 32,078.45                    | 31,699.96                                 |
| Nevada         | 1.00791                         | 1.34514                | 0.98639     | 0.73330     | 33,406.04                    | 33,143.77                                 |
| Nebraska       | 1.00139                         | 1.27263                | 0.97294     | 0.76451     | 28,394.98                    | 28,345.55                                 |
| Oregon         | 0.99935                         | 1.29799                | 0.94603     | 0.72884     | 30,699.03                    | 30,718.85                                 |
| Kansas         | 0.99586                         | 1.32699                | 0.99353     | 0.74871     | 30,070.70                    | 30,195.82                                 |

| States        | Mean Income Ratios 1990/1980 | CV Ratios 1990/1980 | CVs 1990 | CVs 1980 | Mean Incomes (\$) 1990 | Mean Incomes (\$) 1980 <sup>a</sup> |
|---------------|------------------------------|---------------------|----------|----------|------------------------|-------------------------------------|
| Arkansas      | 0.99501                      | 1.31205             | 1.08324  | 0.82561  | 23,662.03              | 23,780.77                           |
| Missouri      | 0.99500                      | 1.33685             | 1.02066  | 0.76348  | 28,913.93              | 29,059.35                           |
| Indiana       | 0.99496                      | 1.32216             | 0.92896  | 0.70261  | 31,214.10              | 31,372.15                           |
| Kentucky      | 0.99101                      | 1.28699             | 1.03860  | 0.80700  | 26,216.66              | 26,454.53                           |
| Mississippi   | 0.98173                      | 1.30123             | 1.09936  | 0.84486  | 23,494.57              | 23,931.71                           |
| Vermont       | 0.98003                      | 1.43785             | 1.03311  | 0.71851  | 26,628.33              | 27,170.83                           |
| Utah          | 0.96719                      | 1.32352             | 0.90398  | 0.68301  | 30,614.17              | 31,652.72                           |
| Wisconsin     | 0.96572                      | 1.36588             | 0.94710  | 0.69340  | 30,616.95              | 31,703.69                           |
| North Dakota  | 0.96421                      | 1.33450             | 1.02080  | 0.76493  | 24,639.44              | 25,553.97                           |
| Iowa          | 0.96403                      | 1.27720             | 0.93219  | 0.72987  | 29,137.47              | 30,224.61                           |
| Michigan      | 0.96356                      | 1.38894             | 0.98505  | 0.70921  | 33,010.67              | 34,259.01                           |
| South Dakota  | 0.95386                      | 1.23545             | 0.96767  | 0.78325  | 23,953.19              | 25,111.78                           |
| Idaho         | 0.94712                      | 1.39847             | 1.04358  | 0.74623  | 26,744.90              | 28,237.99                           |
| Texas         | 0.93975                      | 1.42437             | 1.09316  | 0.76747  | 29,544.71              | 31,438.90                           |
| Colorado      | 0.93958                      | 1.37560             | 0.99680  | 0.72463  | 31,441.86              | 33,463.73                           |
| New Mexico    | 0.93117                      | 1.38685             | 1.06145  | 0.76537  | 25,626.48              | 27,520.74                           |
| Arizona       | 0.91589                      | 1.49504             | 1.10437  | 0.73869  | 28,095.19              | 30,675.22                           |
| Oklahoma      | 0.90177                      | 1.38803             | 1.09976  | 0.79232  | 25,295.50              | 28,050.94                           |
| West Virginia | 0.86726                      | 1.39491             | 1.07543  | 0.77097  | 23,083.16              | 26,616.35                           |
| Louisiana     | 0.86453                      | 1.40019             | 1.14680  | 0.81903  | 24,685.20              | 28,553.33                           |
| Montana       | 0.84494                      | 1.34733             | 1.02708  | 0.76231  | 23,492.22              | 27,803.44                           |
| Alaska        | 0.83841                      | 1.40086             | 0.94956  | 0.67784  | 39,065.86              | 46,594.93                           |
| Wyoming       | 0.76151                      | 1.46890             | 1.00144  | 0.68176  | 26,176.89              | 34,375.15                           |

<sup>a</sup> 1990 Constant dollars.

Source: Estimated by the authors.



**Table E2. Means of Household Income and Coefficients of Variation in the U.S. by Race/Ethnic Groups, 1980-1990**

| <b>Race/Ethnic Groups</b> | <b>Mean Income Ratios</b> | <b>CV Ratios</b> | <b>CVs 1990</b> | <b>CVs 1980</b> | <b>Mean Incomes (\$) 1990</b> | <b>Mean Incomes (\$) 1980<sup>a</sup></b> |
|---------------------------|---------------------------|------------------|-----------------|-----------------|-------------------------------|---|
| Total Population          | 1.03065                   | 1.0770           | 0.81476         | 0.75650         | 36,574.71                     | 35,487.04                                 |
| White                     | 1.02976                   | 1.0804           | 0.79494         | 0.73582         | 38,012.20                     | 36,913.53                                 |
| African American          | 0.97741                   | 1.0371           | 0.90154         | 0.86929         | 24,667.12                     | 25,237.18                                 |
| Native American           | 1.11144                   | 1.0253           | 0.82344         | 0.80313         | 27,910.16                     | 25,111.61                                 |
| Asian                     | 1.08345                   | 1.0150           | 0.75552         | 0.74434         | 44,667.65                     | 41,227.37                                 |
| Hispanic                  | 0.95153                   | 1.0641           | 0.83340         | 0.78320         | 27,760.34                     | 29,174.33                                 |

<sup>a</sup> 1990 Constant dollars.

Note: Data for 1980 were extracted from PUMS 1980. Data for 1990 were extracted from CPS 1990/1991.

Source: Estimated by the authors.

## References

- Bartlett, Robin L., and Timothy I. Miller. "Executive Compensation: Female Executives and Networking." *American Economic Review* 75(2)(1985):266-70.
- Bernheim, B.D., and O. Stark. "Altruism Within the Family Reconsidered: Do Nice Guys Finish Last?" *American Economic Review* 78(1988):1034-45.
- Bernstein, Aaron. "Inequality: How the Growing Gap Between Rich and Poor in America Is Hurting the Economy." *Business Week*, August 15, 1994, pp. 78-82.
- Bogardus, E.S. "Social Distance and Its Origins." *Journal of Applied Sociology* 9(1924-25):216-26.
- Bokemeier, J. "Rediscovering Families and Households: Restructuring Rural Society and Rural Sociology." Presidential Address for the 1996 Annual Meeting of the Rural Sociological Society, Mimeo, 1996.
- Bourdieu, P. "The Forms of Capital." *Handbook of Theory and Research for the Sociology of Education*, edited by J.G. Richardson. New York: Greenwood Press, 1986, pp. 241-58.
- Bradshaw, Y.W., and M. Wallace. *Global Inequalities*. Thousand Oaks, Calif.: Pine Forge Press, 1996, p. 28.
- Bradsher, Keith. "Gap in Wealth in U.S. Called Widest in West." *New York Times*, April 17, 1995, pp. C1, C4.
- Calonius, E. "Blood and Money." *Newsweek*. Special Edition, The 21<sup>st</sup> Century Family, Winter/Spring 1990, p. 82. (Source of numbers used in the article: *Family Business Magazine*, "Buchholz and Crane's Corporate Bloodlines.")
- Clark, John. "Shifting Engagements: Lessons from the 'Bowling Alone' Debate." *Hudson Briefing Paper*, No. 196, October 1996.
- Coleman, J.S. "Social Capital in the Creation of Human Capital." *American Journal of Sociology*, Supplement 94(1988):S95-S120.
- Coleman, J.S., and T. Hoffer. *Public and Private High Schools: The Impact of Communities*. New York: Basic Books, 1987.
- Deininger, K., and L. Squire. "Economic Growth and Income Inequality: Reexamining the Links." *Finance and Development*, March 1997, pp. 38-41.
- Evans, P. "Government Action, Social Capital and Development: Review the Evidence on Synergy." *World Development* 24(1996):1119-32.
- Fields, G.S. *The State of Development Economics: Progress and Perspective*, p. 462.
- Flora, C.B., and J.L. Flora. "The Past and Future: Social Contract, Social Policy and Social Capital." *Increasing Understanding of Public Problems and Policies-1995*, edited by S.A. Halbrook and C.E. Merry. Oak Brook, IL: Farm Foundation, January 1996, pp. 53-64.



- Fox, J. "How Does Society Thicken? The Political Construction of Social Capital in Rural Mexico." *World Development* 24(1996):1089-1103.
- Fukuyama, F. *Trust: The Social Virtues and the Creation of Prosperity*. New York: The Free Press, 1995.
- Gilliland, C.E. "How to Investigate Rural Land's Market Value." *Farm Land Realtor* 37(1985):19-21.
- Granovetter, Mark S. *Getting a Job: A Study of Contacts and Careers*. Cambridge, MA: Harvard Press, 1974.
- Gwilliam, K.R. "Farmland Leasing and Contract Choice in Michigan: The Influence of Social Distance." Ph.D. thesis, Michigan State University, 1993.
- Hyden, G. "The Role of Social Capital in African Development: Illustrations from Tanzania." Paper presented to the Seminar on "Improved Resource Management: The Role of the State versus That of the Local Community," Jyllinge, Denmark, 1-4 November, 1993.
- Johnson, B.B., L. Janssen, M. Lundeen, and J.D. Aiken. "Agricultural Land Leasing and Rental Market Characteristics: Case Study of South Dakota and Nebraska." Unpublished completion report, USDA, Washington, D.C., 1987.
- Kaldor, N. "Capital Accumulation and Economic Growth." In *Further Essays on Economic Theory*, edited by N. Kaldor. New York: Holmes and Meier Publishers, Inc., 1978.
- Korten, D.C. *When Corporations Rule the World*. Kumarian Press, Inc. and Berrett-Koehler Publishers, Inc., 1995, p. 20.
- Kuznets, S. "Economic Growth and Income Inequality." *American Economic Review* 65(1955):1-28.
- Lopez, E.S. "Social Capital and the Educational Performance of Latino and Non-Latino Youth." *JSRI Research Report No. 11*. East Lansing, Michigan: The Julian Samora Research Institute, Michigan State University, 1996.
- McLanahan, Sara, and Karen Booth. "Mother-Only Families: Problems, Prospects, and Politics." *Journal of Marriage and the Family* 51:557-80.
- Meyer, J. "Two-Moment Decision Models and Expected Utility Maximization." *American Economic Review* 77(1987):421-30.
- Nelton, S. "We Could Use a Few Good Numbers." *Nation's Business*, April 1990, p. 53.
- Pope, R.D., R.A. Kramer, R.D. Green, and B.D. Gardner. "An Evaluation of Econometric Models of U.S. Farmland Prices." *Western Journal of Agricultural Economics* 4(1979):107-19.
- Portes, A., and P. Landolt. "The Downside of Social Capital." *The American Prospect*, May-June 1996(26):18-21.
- Putnam, R.D. "Bowling Alone: America's Declining Social Capital." *Journal of Democracy* 6(1995):65-78.

- \_\_\_\_\_. *Making Democracy Work: Civic Traditions in Modern Italy*. Princeton, NJ: Princeton University Press, 1993.
- Robinson, S. "A Note on the U-Hypothesis." *American Economic Review* 66(3):437-40.
- Robison, L.J., and S.D. Hanson. "Social Capital and Economic Cooperation." *Journal of Agricultural and Applied Economics* 27(1)(July 1995):43-58.
- \_\_\_\_\_. "Social Capital and Risk Responses." *Staff Paper No. 96-90*. East Lansing, Michigan: Department of Agricultural Economics, Michigan State University, October 1996.
- Robison, L.J., and A.A. Schmid. "Interpersonal Relationships and Preferences: Evidences and Implications." In *Handbook of Behavioral Economics*, edited by R. Frantz and H. Singh, Vol. 2B. Greenwich, CT: J.A.I. Press, 1991, pp. 347-58.
- \_\_\_\_\_. "Can Agriculture Prosper Without Increased Social Capital?" *Choices*, Fourth Quarter 1994, pp. 29-31.
- Schmid, A.A. Unpublished correspondence on the nature of externalities, October 1996.
- Schmid, A.A., and L.J. Robison. "Social Capital as a Factor of Production." *Journal of Agricultural and Applied Economics* 27(1)(July 1995):59-66.
- Siles, M.E., S.D. Hanson, and L.J. Robison. "Socio-Economics and the Probability of Loan Approval." *Review of Agricultural Economics* 16(1994):363-72.
- Siles, M.E. and L.J. Robison. "Data Bank: Social Capital and Household Income Distributions." Unpublished manuscript, September 1997.
- Smith, A. *The Theory of Moral Sentiments*. (First published, 1759.) Reprints of Economic Classics. New York, NY: Augustus M. Kelley Publishers, 1966.
- Swedberg, R. *Economics and Sociology*. Princeton, NJ: Princeton University Press, 1990.
- Tetsch, K. "Philanthropic Giving Rose 6.2% in '91, Study Says." *New York Times*, June 24, 1992.
- U.S. Department of Commerce, Bureau of the Census. *1990 Census of Population and Housing, Public Use Microdata Sample (PUMS)*. CD90-PUMSAA, August 1993.
- U.S. Department of Labor, Bureau of Labor Statistics. "Jobseeking Methods Used by American Workers." *Bulletin No. 1886*, 1975.
- Whitehead, B.D. "Dan Quayle Was Right." *The Atlantic* 271(April 1993):47-84.
- Williamson, Jeffrey G. "Globalization and Inequality, Past and Present." *The World Bank Research Observer*, Vol. 12, No. 2 (August 1997), pp. 117-35.



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