

Risk Protection, Service Use, and Health Outcomes under Colombia's Health Insurance Program for the Poor

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Unexpected medical care spending imposes considerable financial risk on developing country households. Based on managed care models of health insurance in wealthy countries, Colombia's Régimen Subsidiado is a publicly financed insurance program targeted to the poor, aiming both to provide risk protection and to promote allocative efficiency in the use of medical care. Using a "fuzzy" regression discontinuity design, we find that the program has shielded the poor from some financial risk while increasing the use of traditionally under-utilized preventive services – with measurable health gains.

In developing countries, the inability to smooth consumption directly reduces welfare and leads to informal risk management strategies that stifle productive activity (Paxson 1993, Townsend 1994, Morduch 1995). Because unexpected

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illness is a leading source of economic risk, the expansion of health insurance is therefore a public policy priority in many parts of the developing world (Gertler and Gruber 2002, GTZ, WHO, and ILO 2005, WHO 2010, Mohanan 2012).¹ Such initiatives are often large, centrally-planned programs operated exclusively through the public sector – and they focus primarily on reducing the out-of-pocket price of medical care.²

An early exception is Colombia's *Régimen Subsidiado* (or “Subsidized Regime,” henceforth “SR”). Introduced in 1993, the SR is a pluralistic, publicly-financed health insurance program targeted to the poor.³ Colombians meeting a proxy means-test (determined by the *Sistema de Identificación de Beneficiarios*, or SISBEN) are fully-subsidized to purchase insurance from private, government-approved insurers. In contrast to the classical ‘managed competition’ model of insurance (Enthoven 1978a and 1978b), participating insurers must offer standardized benefits packages and accept standardized premiums. Insurers can, however, form restrictive medical care networks, deny reimbursement for services deemed ‘unnecessary,’ and pay health care providers in ways that encourage higher quality and lower cost medical care (through capitated payment contracts, for example – fixed payments per enrollee per month). Overall, Colombia's SR

¹ For example, one study finds that 5% of Latin American households spend 40% or more of ‘non-subsistence’ income on medical care each year (Xu et al. 2003). As Gertler and Gruber (2002) note, there are two major costs of illness: medical care costs and reduced labor income. Health insurance (our focus) addresses the former, while disability insurance addresses the latter.

² See Abel-Smith (1992); Dow et al. (1997); WHO (2000); WHO Commission on Macroeconomics and Health (2001); Gertler and Solon (2002); Dow, Gonzalez, and Rosero-Bixby (2003); Dow and Schmeer (2003); Gakidou et al. (2006); Pauly et al. (2006); Hughes and Leethongdee (2007); Wagstaff (2007); Wagstaff and Yu (2007); O'Donnell et al. (2008); Pauly, Blavin, and Meghan (2008); and Gruber, Hendren, and Townsend (2012).

³ According to Article 153 of Law 100 (authorizing the creation of the SR), one of the SR's guiding principles is *Proteccion Integral*: “The System of Social Security in health will provide health care to the population in: education, information, health promotion and prevention, diagnostic, treatment and rehabilitation in quantity and quality according to the *Plan Obligatorio de Salud*.”

shares features of managed care models of health insurance that emphasize cost-containment and allocative efficiency traditionally found only in wealthy countries. It therefore represents important early experience in a markedly different institutional environment.

This paper studies the impact of the SR on financial risk protection, service use, and health outcomes among Colombia's poor. Program eligibility is supposed to be determined according to a discrete threshold in the continuous SISBEN index, so in principle we could use a regression discontinuity design to do so. Because SISBEN scores are manipulated in practice (BDO and CCRP 2000, DNP 2001, 2003a, and 2003b, Fresneda 2003, Camacho and Conover 2011), we instead use underlying index components collected through independent household surveys to generate our own (un-manipulated) SISBEN score calculations. We then instrument for SR enrollment with our re-constructed eligibility measure (Hahn, Todd, and Van der Klaauw 2001). In general, our estimates are robust across a variety of parametric and non-parametric specifications.

Despite our strategy for addressing manipulation of eligibility, our approach has limitations. First, because we use household surveys to implement a "fuzzy" regression discontinuity design, our samples are relatively small. Second, we measure SISBEN index components after official SISBEN classification occurred, presumably resulting in a degree of measurement error. Third, due to financial shortfalls, many of Colombia's *municipios* (hereafter, "counties") used eligibility thresholds that fell short of the official one. Following Chay, McEwan, and Urquiola (2005), we therefore estimate and use county-specific thresholds. These limitations introduce noise into the regression discontinuity design and generally bias us against finding behavioral responses to the SR.

We first find evidence that by the mid-2000s, the SR succeeded in protecting poor Colombians from financial risk associated with the medical costs of unexpected illness. In particular, SR enrollment appears to have reduced the

variability of out-of-pocket spending for inpatient care. Despite this reduction in risk, however, we observe little evidence of meaningful portfolio choice effects (changes in the composition of household assets, human capital investments, or household consumption), perhaps because the SR falls short of providing full insurance.

Our results also suggest that SR enrollment is associated with large increases in the use of traditionally under-utilized preventive services – some of which nearly doubled. Moreover, we find evidence of health improvement under the SR as well – specifically, gains along margins sensitive to the increases in preventive care that we observe. There is more mixed evidence of changes in the use of curative services (although theoretical predictions about the use of curative care are ambiguous).

We conclude by discussing the underlying behavioral mechanisms that may explain our results. Because the SR is complex and multi-faceted, it is important to note that we cannot draw firm inferences about them; we emphasize this as an important direction for future research. Overall, we highlight two mechanisms that we suspect are important: high-powered supply-side incentives and the possibility that enrollees receive care from higher-quality private sector facilities.

I. Background and Policy Context

A. Public Sector Health Insurance for Colombia's Poor Prior to the Reform

Prior to the introduction of the SR in 1993, roughly 25% of Colombians (a subset of those with formal sector jobs) had any form of explicit health insurance (Pinto 2008). However, Colombians lacking formal insurance also had a degree of implicit insurance provided through the public sector. Specifically, they could receive medical care from public sector hospitals and clinics for a fraction of the full cost of their services; out-of-pocket payments were generally progressive and

loosely based on socio-economic status. Public sector facilities, in turn, covered their losses with direct transfers from national and local governments, and health care professionals were typically paid fixed salaries that did not reward productivity. Thus, poor Colombians effectively had a degree of implicit health insurance coupled with inefficient provider incentives – and the reform that we study aimed to expand coverage while improving efficiency.

B. Overview of Colombia's Subsidized Health Insurance Regime for the Poor

Under Law 100 in 1993, Colombia introduced the SR, a novel form of publicly-financed health insurance for the poor (Gwatkin, Wagstaff and Yazbeck 2005, Escobar 2005). Primarily through SR expansion, formal health insurance coverage in Colombia grew from about 25% of the population in 1993 to 80% in 2007 (CENDEX 2008). The SR is organized as a variant of the classical 'managed competition' model (Enthoven 1978a and 1978b). Beneficiaries are fully subsidized to purchase health insurance from competing health plans. During our study period (the mid-2000s), subsidies were financed by a combination of public resources including payroll taxes and national and local general revenue. These resources are transferred to county governments, which in turn are responsible for eligibility determination, enrollment, and contracting with health plans. This role of local government has important implications for our empirical analysis.

Health plans charge county health authorities a government-regulated premium per enrollee and offer a standardized package of benefits (see Appendix A for the details of these benefits).⁴ Participating health plans then act as group purchasers of health services for their enrollees by contracting with a network of health

⁴ During our study period, the benefits package of the SR (*Plan Obligatorio de Salud Subsidiado*) emphasized coverage for primary and basic outpatient services, drugs, and some catastrophic care. There was limited coverage for specialist services, and there were substantial gaps in coverage for hospital care.

facilities and clinicians. Because premiums and benefit packages are standardized by law (unlike the classical ‘managed competition’ model), health plans compete for enrollees on the basis of provider networks and service quality. In practice, however, very few cities had more than one insurer during the years that we study.

Insurers’ ability to contract with health care providers (hospitals and medical groups) for more efficient service provision is a central innovation of the SR. Insurers receive premiums (fixed payments per enrollee per unit time) for all covered services, giving them strong incentives to constrain total spending. Insurers then (partially) transmit these incentives to provider organizations through capitated payment contracts, and they have the authority to deny reimbursement for services deemed inefficient.⁵

In the next section we describe the major demand- and supply-side incentive changes under the SR, and Section II then provides predictions based on them for health service use, out-of-pocket spending on medical care, and health outcomes.

C. Changes in Demand- and Supply-Side Incentives under the SR

There were important changes in both demand- and supply-side incentives with the introduction of the SR. On the demand-side, out-of-pocket prices for covered curative services are 10% of the full price (as opposed to 30% for poor uninsured Colombians).⁶ Moreover, total out-of-pocket spending per episode of illness each

⁵ Provider organizations, in turn, must transmit the incentives they face to individual clinicians whom they employ. Our interviews with stakeholders in the Colombian health care system suggest that organizations solve this agency problem through non-financial rather than payment-based incentives (systematic data on organizational incentives and clinician contracts is unfortunately unavailable).

⁶ The Colombian Ministry of Social Protection maintains a fee schedule (*SOAT*) regulating out-of-pocket prices. For example, the full price for a physician consultation was 17,300 pesos in 2007 and 18,460 pesos in 2008 (approximately US \$10). Medical facility social workers assess uninsured patients’ “ability to pay” and may charge them sliding scale out-of-pocket fees (greater than 10% of the full price) on a case-by-case basis.

year is capped at half of the monthly minimum wage for SR beneficiaries – while the same cap is six times as large for uninsured patients (*Decreto 2357* and *Acuerdo 260* of the *Consejo Nacional de Seguridad Social en Salud*). Notably, preventive services are essentially free regardless of insurance status.

On the supply-side, changes in provider payment contracts created substantially different incentives for the provision of health services. Specifically, there are two types of contracts between insurers and provider organizations under the SR: capitated primary care contracts and fee-for-service specialty care contracts. For primary care, insurers pay providers fixed amounts per month for all services used by enrollees (“capitation”). These contracts create strong incentives for providers to constrain total spending on primary care and have important implications for both preventive and curative health care (as discussed in the next section). Importantly, promotion of preventive care can be a central means of reducing total expenditures.⁷

For specialty care, insurers generally pay providers a pre-determined fee for each covered service that they supply (i.e., on a “fee-for-service” basis). These contracts encourage the provision of all reimbursable services (both efficient and inefficient). However, SR insurers also have the authority to deny reimbursement on a case-by-case basis for inefficient specialty care (termed “utilization review”), allowing them to limit wasteful service use.⁸

D. Eligibility for the SR

Eligibility for the SR is determined using a poverty-targeting index called SISBEN (or *Sistema de Identificación de Beneficiarios*). The original SISBEN

⁷ Many medical professionals believe that increasing prevention reduces overall medical spending – whether or not this is true depends on prices, the health production function, and competing risks. Cohen, Neumann and Weinstein (2008) find that 20% of preventive measure in the US are cost saving.

⁸ Utilization review does little to promote services traditionally used sub-optimally.

index consisted of fourteen components measuring different aspects of household well-being (such as housing material, access to public utilities, ownership of durable assets, demographic composition, educational attainment, and labor force participation – for a complete description, see Appendix B).⁹ On each dimension, households are classified according to mutually exclusive, collectively exhaustive categories with varying weights assigned to each category; these weights vary between urban and rural areas. A household's SISBEN score is then calculated by summing points across components. Possible scores range from 0 to 100 (with 0 being the most impoverished) and are divided into six strata. Households scoring in SISBEN strata 1 and 2 (the lowest strata) are eligible for the SR (below 48 in urban areas, below 31 in rural areas).¹⁰

E. Eligibility and Enrollment in Practice

Although eligibility for the SR increases the likelihood of enrollment, neither one necessarily implies the other for at least three reasons: misclassification or manipulation of SISBEN scores, shortfalls in local government revenue, and enrollment that preceded SISBEN enumeration.¹¹

First, both local governments and households have incentives to manipulate SISBEN scores. Local governments receive fixed transfers from the national government for each resident they enroll, creating incentives to maximize enrollment. The selective enrollment of key constituents can also provide political

⁹ Eligibility also varies within households because certain demographic groups (including pregnant women and young children) are prioritized for enrollment.

¹⁰ SISBEN eligibility shifts abruptly at each county's *cabecera* boundary, an administrative demarcation formally distinguishing urban and rural parts of each county and loosely corresponding to the fringe of public utility infrastructure. Distinct urban and rural SISBEN scales are applied to households on corresponding sides of the boundary, differing both in component parts and in the weighting of response categories for each component. We implemented a research design exploiting these urban/rural index differences, but inconsistent application of the rural index and data limitations prevent us from drawing meaningful conclusions from it. In this paper we therefore focus on urban eligibility.

¹¹ Administrative mistakes in the enrollment process are also important.

benefits (Camacho and Conover 2011). Households prefer enrollment over “uninsurance” as well because co-payments are lower for SR beneficiaries than for those lacking formal insurance. Consistent with both types of incentives, there is evidence of considerable SISBEN score manipulation between 1997 and 2003 (Camacho and Conover 2011).¹²

Second, most local governments lack sufficient revenue to finance the enrollment of all eligible residents. According to law, those with lower SISBEN scores and those belonging to specific targeted groups (such as children under five and pregnant women) are therefore prioritized for enrollment.¹³ This means that many counties use *de facto* eligibility thresholds that fall below the uniform national threshold.

Third, some counties began enrolling residents in the SR before all of their residents had been classified using SISBEN. These counties instead used other means-test criteria such as residents’ *estrato*, a neighbourhood-level measure of socio-economic status used to establish electricity prices paid by local households.

In general, these realities of the program’s implementation have two broad implications for our empirical analyses (which we discuss in greater detail in Section IIIB). One is the necessity of an empirical strategy that addresses manipulation of SISBEN scores. The other is that *de facto* county threshold must be estimated (because they are not officially reported).

¹² Using results from the 2005 population census, the Colombian newspaper *El Tiempo* reports that there are more SR enrollees than residents in some counties (*El Tiempo*, October 26, 2006). Camacho and Conover (2011) show that the distribution of official SISBEN scores exhibits both large leftward shifts in density over time and the formation of a mass point just to the left of the national eligibility threshold in urban areas. Neither are present in Colombian household surveys. The former suggests misrepresentation by households, while the latter suggests misrepresentation by enumerators or officials.

¹³ The laws formalizing this prioritization are *Acuerdos 244* and *253* of the *Consejo Nacional de Seguridad Social en Salud*. This prioritization also means that although SISBEN scores are calculated at the level of family “nucleus,” individuals within families can vary in enrollment status; we observe this in our household survey data.

F. Previous Studies

Several previous empirical papers have studied Colombia's SR as well (see Giedion and Uribe (2009) for a comprehensive review). Trujillo, Portillo, and Vernon (2005) use propensity score matching to compare SR enrollees to observationally similar uninsured Colombians, finding that the SR is associated with greater medical care use. Giedion et al. (2009) also use propensity score matching and find that SR enrollment is associated with more visits to health care providers, higher vaccination coverage rates, and fewer reports of not seeking medical care for financial reasons. Gaviria, Medina, and Mejía (2007) instrument for SR enrollment using length of time a household head lived in his/her current place of residence and report that SR enrollment is associated with better self-reported health, more preventive and curative outpatient care, and fewer hospitalizations. Finally, matching official SISBEN score data with birth records, Camacho and Conover (2008) use a regression discontinuity design in two samples: a county lacking clear evidence of manipulation and selected counties observed shortly after SR implementation. They find that SR enrollment is associated with increased birth weight and better APGAR scores but not antenatal care use, medical supervision of deliveries, or probability of hospital delivery. Relative to previous studies of the SR, our paper employs a different strategy for addressing manipulation of program eligibility, uses an approach that disentangles the role of the SR from other public programs using SISBEN, and studies a broader range of behavioral responses to health insurance.¹⁴

¹⁴ See Card, Dobkin, and Maestas (2008), Card, Dobkin, and Maestas (2009), Chandra, Gruber, and McKnight (2010a), and Anderson, Dobkin, and Gross (2012) for recent regression discontinuity studies of health insurance in the United States. For recent work on the United States using other methodologies, see Doyle (2005), Dafny and Gruber (2005), Finkelstein (2007), Finkelstein and McKnight (2008), Kolstad and Kowalski (2010), Chandra, Gruber, and McKnight (2010b), and Kowalski (2011).

II. Empirical Predictions on the Interaction of Supply- and Demand-Side Incentives

Although we emphasize the centrality of changes in supply-side incentives under the SR, out-of-pocket prices for covered services also fell. This section therefore considers the joint effect of supply- and demand-side changes in formulating predictions about changes in risk protection, the use of preventive and curative services, and health outcomes. Section IV then provides empirical evidence on each.

Risk Protection and Out-of-Pocket Medical Spending.— Protecting households against medical care costs associated with unexpected illness is a central objective of health insurance. As Appendix A shows, the SR covers many of the most expensive services including trauma care, dialysis, and major joint replacement (hips and knees, for example). It also caps total out-of-pocket spending per episode of illness each year at half of the monthly minimum wage. The SR should therefore reduce the variability of out-of-pocket medical spending. Assuming that the price elasticity of demand for medical care is less than one, the SR should also reduce total out-of-pocket spending for medical care.

Preventive Health Services.— Most preventive services in Colombia are free regardless of insurance status (i.e., free both for SR enrollees and the uninsured), so the main difference in incentives for preventive care originate on the supply-side. Primary care providers are paid on a capitated basis, so they have strong incentives to limit total primary care spending. Increasing preventive service use can reduce the need for more costly curative care – and can therefore be cost-

saving – so capitated primary care incentives are likely to increase preventive care use.¹⁵

An opposing force at work is *ex ante* moral hazard: SR enrollees may have weaker incentives to use preventive services because they pay less for curative care (Nordquist and Wu 1976, Phelps 1978). Although we fail to find evidence of *ex ante* moral hazard (as shown in Section IVE) and there is little evidence of it in other studies (Kenkel 2000), our preventive care estimates reflect the net effect of these forces.¹⁶ Because preventive services generate important positive externalities (both pecuniary and infectious disease-related), increases in their use presumably improve welfare.¹⁷

Curative Medical Care.— Three changes under the SR are likely to influence the use of curative medical care: reductions in out-of-pocket prices, increases in the use of preventive services, and supply-side incentives for limiting total medical spending. First, for negative price elasticities of demand, reductions in out-of-pocket prices will increase the use of curative care. This increase is inefficient under standard assumptions (*ex post* moral hazard); however, curative care in developing countries may produce positive externalities, and there may also be important credit constraints. Second, any increases in preventive service use may reduce the use of curative care (an efficient result). Third, both capitation and utilization review produce incentives for providers to limit the use of curative

¹⁵ One randomized controlled trial in the US and another in the UK show that capitation increases preventive service use (Manning et al. 1984, Lennon et al. 1990). This finding is also consistent with the evidence from observational studies summarized by Miller and Luft (1994) and more recent research by Keenan et al. (2009).

¹⁶ The RAND Health Insurance Experiment reports no sizeable or significant effect of insurance coverage on health behaviors (smoking, drinking, and exercise) (Newhouse and the Insurance Experiment Group 1993). Medicare coverage has also been reported not to increase unhealthy behaviors (Dave and Kaestner 2006).

¹⁷ Popular conditional cash transfer programs – including the Familias en Acción program in Colombia – aim to increase the use of preventive services even though they are otherwise available for free. At least eleven developing countries have introduced such conditional cash transfer programs (Fiszbein and Schady 2009).

medical care. Some reductions may be efficient (counterbalancing demand-side incentives leading to *ex post* moral hazard) while others may be inefficient (leading to “stinting” – see for example Ellis and McGuire (1990), Holmstrom and Milgrom (1991), Ma and McGuire (1997), and McGuire (2000)). Predictions about changes in curative care under the SR as well as their welfare implications are therefore ambiguous.

Health Outcomes.— Both preventive and curative services are inputs into health production. If at least one of these two types of care increases and neither of them decreases, health should presumably improve to some degree. Alternatively, the prediction is ambiguous if one type increases and the other decreases. In our empirical analyses, we are able to examine the prevalence of infectious diseases common among children (diarrhea, cough and fever) as well as the number of days lost to illness.

III. Data and Empirical Strategy

A. Data

Our empirical approach requires household survey data containing three types of information: (1) enrollment in the SR, (2) components of the SISBEN index (enabling us to simulate SR eligibility), and (3) potential behavioral responses and outcomes of interest (both welfare-improving and distortionary). There are two candidate Colombian household surveys that meet these criteria: the *Encuestas de Calidad de Vida* (ECV) and the Demographic and Health Surveys (DHS). The ECVs are nationally-representative household surveys designed to measure socio-economic well-being and “quality of life,” broadly defined. The DHS data reports detailed fertility, health, and socio-economic information for nationally-representative samples of fertile age women (defined as ages 15-49) and their

households. Because the *de facto* implementation of the SR occurred in 1996/1997, we use the 2003 ECV and the 2005 DHS for our analyses.¹⁸ Table 1 shows descriptive statistics by type of behavior/outcome for the full samples as well as those with and without SR coverage.

As our empirical strategy requires, we calculate household-level SISBEN scores to simulate SR eligibility because simulated eligibility should not reflect manipulation of SISBEN scores (Camacho and Conover 2011).¹⁹ However, not all household surveys contain all necessary components of the SISBEN index. Appendix B provides a complete description of the SISBEN components present in each survey.²⁰ We impute values using ordered probit models for the few variables that are missing (firm size and per-capita income in the DHS and primary roof material in both surveys). *Estrato* is also missing for some households and we impute them on the basis of the most common in the household's neighborhood.

B. Empirical Strategy

Instrumenting for Enrollment with Simulated Eligibility.—In principle, the SISBEN index's SR eligibility threshold can be used to study behavioral responses associated with SR enrollment. This discontinuity induces an abrupt shift in eligibility (and enrollment) along otherwise smooth distributions of household characteristics; coincident shifts in behaviors and outcomes can reasonably be linked to the program. However, adverse or propitious selection into eligibility – or manipulation according to unobserved household

¹⁸ There was also a Colombian DHS survey conducted in 2000, but it is much smaller and contains few outcome variables of interest. We do not use the 1997 wave of the ECV because SR enrollment was still very low in that year.

¹⁹ Reliable official SISBEN scores are also not available in the datasets that we use.

²⁰ In theory, SISBEN scores should be calculated at the family (or “nucleus”) level. However, we treat entire households as families given reports that SISBEN enumerators adopted this definition in practice due to difficulties in conforming to the technical definition.

characteristics, as discussed in Section IE – is likely to bias the estimates of interest (McCrary 2008).

To circumvent this difficulty, we employ an instrumental variables strategy closely resembling one proposed by Hahn, Todd, and Van der Klaauw (2001). Conceptually, we seek to reconstruct ‘true’ SISBEN scores when both official SISBEN scores and observed SR enrollment reflect manipulation. To do so, we calculate SISBEN scores for each household in the ECV and DHS data and then use calculated scores to instrument for SR enrollment (for prominent examples of simulated instruments, see Currie and Gruber (1996a and 1996b), and Cutler and Gruber (1996), and Hoxby (2001)).²¹

Using urban households with simulated SISBEN scores near the urban eligibility threshold (we do not use rural households because of sufficiently inconsistent application of the rural scale)²², we could in principle begin by estimating the following first-stage equation for individuals i in household h :

$$(1) \quad enroll_{ih} = \alpha + \gamma below_h + \beta SISBEN_h + \sum_k \delta_k estrato_{hk} + \varepsilon_{ih},$$

where *enroll* is an indicator for whether or not individual i is enrolled in the SR, *below* is an indicator for simulated SISBEN score lying below the eligibility threshold, *SISBEN* is simulated SISBEN score, and *estrato* is a dummy variable for an *estrato* category (a neighborhood-level measure of socio-economic status used to establish electricity prices paid by local households). Using Two-Stage

²¹ We emphasize “old” SISBEN scores – those calculated using the official scale in effect between the beginning of the SR and 2003. Enrollees eligible only under the old scale were not disenrolled with the introduction of the “new scale,” and the old (but not the new) eligibility discontinuity is evident in the 2005 DHS.

²² The combination of smaller sample size and inconsistent application of the scale in rural areas means that our first-stage relationships are considerably weaker than in urban areas. In the DHS, the first-stage F-statistics are about 9 for the sample of children (N=574) and about 3 for the sample of mothers (N=302). In the ECV, the first-stage F-statistics are about 18, but the size of the rural sample is roughly one quarter of the urban one (1318 vs. 4129).

Least Squares (2SLS), we could then estimate the following second-stage equation:

$$(2) \quad outcome_{ih} = \varphi + \lambda enroll_{ih} + \theta SISBEN_h + \sum_k \pi_k estrato_{hk} + \xi_{ih},$$

instrumenting for *enroll* with *below*. The relationship between behavioral outcomes of interest (*outcome*) and SR enrollment would then be captured by estimates of the parameter λ .

Estimating County-Specific Eligibility Thresholds.— As described in Section IE, financial shortfalls led many Colombian counties to use SR eligibility thresholds at SISBEN scores below the official national threshold. We therefore use county-specific eligibility thresholds. In addition to improving the strength of our first stage, this approach offers another key benefit: because some local governments use the official national threshold for other public benefits, changes in outcomes observed at county-specific SR thresholds will not reflect behavioral responses to other public programs. (Section IVA shows that participation in other public programs is not discontinuous at county-specific thresholds.)

Exact county-specific eligibility thresholds are unknown, so we estimate them following Chay, McEwan, and Urquiola (2005). Specifically, using our full samples, we establish county-specific breaks in SR eligibility at the SISBEN score that maximize the goodness-of-fit of a model of SR enrollment as a function of a dichotomous indicator for whether or not a household's score falls below the threshold.²³ This approach establishes thresholds that maximize the percentage of

²³ There are 84 counties in the ECV and 220 counties in the DHS. We exclude individuals from a few counties using two criteria related to having very few observations in some counties. One is having the first percentile in the SISBEN score distribution lie above the national threshold or the 99th percentile score lie below the national threshold. The other is having an estimated threshold with those below it having relatively lower SR enrollment rates than those above it. The total number of observations excluded for these reasons is minor (3.8% of the sample in the ECV data

individuals correctly classified as eligible in each county. Threshold means in the ECV and DHS are 45.03 and 45.83, respectively, and their interquantile ranges are 4.96 and 3.89. Intuitively, estimated thresholds are positively correlated with current county income (the sum of taxes, transfers from the state, and non-tax income) and negatively correlated with the share of households having unmet basic needs.²⁴

We then use county-specific thresholds to re-code the variable *below* for each individual i in households h and Colombian counties c and estimate the following first stage equation:

$$(3) \quad enroll_{ihc} = \alpha + \gamma below_{hc} + \beta SISBEN_h + \phi SISBEN_diff_{hc} + \sum_k \delta_k estrato_{hk} + \mu_c + \varepsilon_{ihc},$$

where *below* is now an indicator for whether or not individual i 's simulated SISBEN score falls below the eligibility threshold in the individual's county c , *SISBEN_diff* is the difference between an individual's simulated SISBEN score and the estimated eligibility threshold in the individual's county (i.e., relative SISBEN score), μ_c represents county fixed effects (allowing us to focus on within-county variation in simulated eligibility across county-specific thresholds), and all other variables are defined as in equation (1). To adhere transparently to the identifying assumption that individuals with simulated SISBEN scores very near the threshold are comparable with the exception of their eligibility, we conservatively focus on individuals whose calculated scores lie within two index

and 5.2% of the sample in the DHS data). Estimated thresholds are constrained to be lower than 49 to allow for measurement error.

²⁴ To assess this, we ran an OLS regression of county-specific eligibility thresholds on current county income and the share of households in the county having unmet basic needs (a composite indicator that collapses across household population density, water source, toilet availability at home, education of household head, ratio of employed household members to all household members, and children's school attendance rate). Using the ECV sample, the estimate for current county income is 0.0091 with a standard error of 0.004; in the DHS, the estimate is 0.0043 with a standard error of 0.002. For the share of households with unmet basic needs, the estimates are -0.0456 and -0.0444 with standard errors of 0.024 and 0.015, respectively.

points of the county-specific cutoff (our main estimates persist across various bandwidths, as shown in Section IVF).²⁵

Figures 1A and 1B use ECV and DHS data to show SR enrollment and “uninsurance” by simulated SISBEN score relative to county-specific eligibility thresholds. Each county’s threshold is normalized to zero, and the figure then shows means and 95% confidence intervals for each SISBEN index integer relative to the threshold as well as non-parametric kernel density plots on either side. The figure illustrates large discrete increases in the probability of enrollment and concomitant decreases in the probability of uninsurance at the threshold ranging between 25 to 30 percentage points. Figures 2A and 2B show the enrollment and “uninsurance” separately by regions of Colombia as well. It is worth noting that SR enrollment falls short of 100% at low (calculated) SISBEN scores. This may not only be due to mistargeting, but also to our use of household survey data (rather than official data) to construct SISBEN scores

Using our re-coded variable *below* to instrument for *enroll*, we then estimate the following equation by 2SLS:

$$(4) \text{ outcome}_{ihc} = \varphi + \lambda \text{enroll}_{ihc} + \theta \text{SISBEN}_h + \psi \text{SISBEN_diff}_{hc} + \sum_k \pi_k \text{estrato}_{hk} + \mu_c + \xi_{ihc},$$

where the estimate of interest is the estimate of λ .²⁶ Section IVF shows that our results are robust across bandwidths to including higher-order relative SISBEN score polynomials, using interactions between relative SISBEN scores and

²⁵ Because eligibility thresholds vary by county, our approach essentially averages across sample respondents with different absolute SISBEN scores. To investigate how our estimates vary with absolute SISBEN score, we also estimate variants of equation (4) below with interactions between SR enrollment and absolute SISBEN score (instrumenting for this term with interactions between an indicator for falling below county-specific thresholds and absolute SISBEN score) and find insignificant estimates for this interaction term (see Section IVG).

²⁶ We estimate equations (3) and (4) using linear models; marginal probabilities computed using bivariate probit models yield similar results to the 2SLS estimates for dichotomous outcomes examined throughout the paper. We calculate our standard errors clustered by county.

simulated eligibility, and excluding county fixed effects. We also estimate λ using local linear regression without any covariates other than relative SISBEN score.²⁷

Limitations of the Empirical Strategy.—Although our approach addresses key irregularities in SR implementation, it also has shortcomings. First, because the ECV and DHS household survey data that we use to calculate SISBEN scores was collected after official SISBEN classification, our calculations presumably include measurement error (even absent manipulation of official SISBEN status). Second, our estimates of county-specific eligibility thresholds presumably do not match those used in practice exactly, introducing additional noise into the relationship between our measure of simulated eligibility and actual enrollment. Third, manipulation of official SISBEN scores further weakens the strength of the first stage relationship. Fourth, our sample sizes are relatively small in the neighborhood of the eligibility thresholds. Overall, these limitations clearly compromise the power of our estimates (few are significant at significance thresholds below $\alpha=0.10$) and suggest that our results should also be interpreted as lower bounds.

IV. Results

This section presents empirical evidence on a variety of important behavioral responses to the SR. We begin by demonstrating that both background characteristics not influenced by the program and participation in other social programs using SISBEN are balanced across estimated SR eligibility thresholds. We then investigate the effectiveness of health insurance in protecting households against financial risk (and possibly changing the optimal composition of household spending and assets). Next, we study changes in the use of traditionally

²⁷ For comparison, we also estimate equation (4) by OLS and show the results at the bottom of our tables.

under-utilized preventive services in light of the SR’s emphasis on allocative efficiency. In doing so, we also examine changes in health outcomes that are sensitive to the use of important preventive services. We then analyze how SR enrollment is related to changes in the use of curative medicine. To investigate possible behavioral distortions, we also test for reductions in private health investments (*ex ante* moral hazard) and insurance ‘crowd-out’ associated with SR enrollment. Finally, we assess the robustness of our estimates through a variety of additional parametric and non-parametric specifications.²⁸

A. Balance across Discontinuous Eligibility Thresholds

Our empirical approach assumes that no individual or household characteristics – other than SR enrollment – that could influence the outcomes of interest vary discontinuously across our estimated eligibility thresholds. To test this assumption, Table 2 shows results obtained by estimating equations (3) and (4) for individual attributes that could not reasonably change in response to SR enrollment (such as age or educational attainment among adults). Consistent with our assumption, the estimates are not generally distinguishable from zero. (Examining the distribution of simulated SISBEN scores across eligibility discontinuities, Appendix C shows evidence in support of this assumption as well.)

We then consider whether or not our estimates from equation (4) could be attributed to participation in other public programs that also use the SISBEN index. Before investigating this possibility directly, we first note that it is unlikely because these programs use the uniform national eligibility threshold,

²⁸ As expected, power is lower in all analyses if we use the uniform national threshold rather than estimated county-specific ones. In contrast to the F statistics shown in our main tables, using the uniform national threshold yields first stage F statics of 4.7 in the ECV and 18 in the DHS. Even so, many of our key results on health care use and health outcomes are nonetheless present when we use the uniform national threshold.

while we estimate and utilize *de facto* county-specific thresholds for the SR – many of which fall below the uniform national threshold. To confirm this, we re-estimate equations (3) and (4) using a dichotomous indicator for participation in these other programs – including job training, home mortgage subsidies, education vouchers, *Hogares Comunitarios* (a large child care program), and services provided by the *Instituto Colombiano de Bienestar Familiar* (the largest social welfare agency in Colombia) – as the dependent variable in equation (4).²⁹ Table 2 presents these results, suggesting that participation in other programs is also balanced across county-specific SR eligibility thresholds.

B. Financial Risk Protection and Portfolio Choice

Although a key innovation of the SR is its emphasis on allocative efficiency, a central objective of any health insurance program is to provide protection against financial risk. We therefore begin by examining the relationship between SR enrollment and both level and variability of out-of-pocket medical spending. To construct our variability measure, we first calculate mean individual spending separately among those enrolled and those not enrolled in the SR. For each individual, we then measure the difference between each individual's spending and the mean among those with the same enrollment status, constructing our variability measure as the absolute value of this difference. We analyze outpatient and inpatient out-of-pocket spending (within the past twelve months) separately

²⁹ The program *Empleo en Acción* (a workfare program active in 2003 but abolished in 2004) is the only public program of which we are aware that used the uniform national SISBEN eligibility threshold for which the 2003 ECV does not contain data. Participation in the program was low (only 7.4% of those in SISBEN strata 1 or 2) and provided no benefits directly related to health (it paid 60% of the minimum wage to individuals who worked on official program projects an average of at least 30 hours per week for 2.4 months) (IFS-SEI-Econometria 2005). We do not have information about participation in *Familias en Acción*, a Colombian conditional cash transfer program, but the eligibility threshold for this program (36) is considerably lower, falling more than two standard deviations below the mean of our estimated SR thresholds.

and emphasize the latter because of idiosyncrasies in how outpatient spending is reported.³⁰

We first graphically examine shifts in the distribution of medical spending associated with simulated eligibility for the SR. Figure 3 shows cumulative density functions for inpatient medical spending in the preceding year separately for those falling above and below county-specific thresholds (using our sample of those within two index points of the cutoff).³¹ Both distributions are heavily right-skewed, but mass in the distribution for those who are eligible (those below the threshold) falls to the left of the distribution for those who are ineligible. No clearly comparable pattern is present for outpatient medical spending (Figure 4). Overall, these figures suggest that SR eligibility is associated with reductions in right-tail inpatient medical spending – a pattern consistent with risk protection.³²

The first four columns of Panel A in Table 3 then present econometric results obtained by estimating equations (3) and (4) for medical spending by type. The first row presents IV estimates for SR enrollment, and the second row reports intent-to-treat (ITT) estimates for simulated SR eligibility (estimates for a dummy variable coding whether or not an individual falls below the eligibility threshold obtained by OLS regressions of outcomes on this dummy and the other covariates

³⁰ The ECV 2003 question about outpatient spending in the preceding 30 days excludes outpatient costs associated with illness ultimately leading to hospitalization; the inpatient expenditure question asks about all inpatient spending in the past 12 months.

³¹ These graphs do not correspond exactly to our formal RD estimates for several reasons. One is that they only include observations with positive expenditures and hence do not capture differential selection into any spending between those above vs. below the eligibility threshold. Another is that using our bandwidth of two sample, they depict mean differences between those above vs. below the eligibility threshold (rather than differences between the two groups as one converges to the threshold from above vs. below). Conducting a Kolmogorov-Smirnov test for equality of distributions, we are unable to reject the null hypothesis of equality, presumably because of these limitations and small sample sizes when conditioning on any spending. They are nonetheless suggestive.

³² These figures may understate the magnitude of risk protection under the SR for two reasons: (1) some people may not receive medical care at all, but nonetheless incur non-medical financial expenses due to poor health, and (2) the figures show densities by eligibility rather than enrollment status (because enrollment is endogenous).

in equation 4). The first column suggests that SR enrollment lowers mean inpatient spending by about 60,000 pesos ($\alpha=0.10$), around a 30% reduction among those using any inpatient services. Perhaps more importantly for assessing protection against financial risk, the third column shows that SR enrollment is associated with reductions in the variability of inpatient medical spending ($\alpha=0.05$).³³ There is no statistically significant association for outpatient care. Appendix D Figure 1 graphically shows all outcomes examined in Panel A across county-specific eligibility thresholds (essentially, graphical versions of our intent-to-treat analyses).

Overall, the results shown in Table 3 Panel A suggest that SR enrollment is associated with meaningful risk protection benefits. By reducing household exposure to financial risk, SR enrollment could also produce meaningful changes in the composition of household assets, human capital investments, and household consumption (i.e., portfolio choice effects). Specifically, it may increase investments not previously undertaken because of costly informal risk-management activities (such as precautionary saving). Panel B of Table 3 presents estimates for durable goods not used to construct the SISBEN index (car and radio ownership) as well as household education and consumption expenditures. In general, it implies that SR enrollment is not associated with discernible portfolio choice effects (perhaps because the SR falls short of providing full insurance).

C. Preventive Service Use and Health Status

As described earlier, a key innovation of the SR is its focus on improving allocative efficiency in medical care. In particular, increasing the use of highly

³³ Although this result is unsurprising given that Colombia's reform caps the amount that SR enrollees are required to pay out-of-pocket, it cannot be taken for granted in a developing country context. For example, Wagstaff and Lindelow (2008) find that health insurance in China increases financial risk due to a combination of poor regulation and asymmetric information.

beneficial preventive services – many of which produce large positive externalities – is likely to improve welfare. These externalities are due both to reduced rates of infectious disease transmission and to reduced curative care costs borne by others through risk pools.

The first two columns of Table 4 report estimates for different types of preventive care use obtained from equations (3) and (4) (Appendix D Figure 2 shows graphical versions of the intent-to-treat analyses). In general, they suggest substantial increases in the use of preventive health care services.³⁴ Specifically, SR enrollment is associated with a 29 percentage point increase in the probability of a preventive physician visit in the past year (a 75% increase; $\alpha=0.01$). Importantly, the SR is also associated with 1.50 more growth-monitoring and well-care visits in the past year, a 50% increase relative to uninsured children ($\alpha=0.05$). These well-care visits are a principal way of addressing important childhood health problems (parents receive nutritional advice, iron supplements, and de-worming medications, for example – which are important for strengthening children’s immune systems and making them less susceptible to major childhood illnesses such as severe diarrhea and acute respiratory infections/pneumonia).³⁵

We then investigate whether or not health status has improved under the SR – in particular, dimensions of health that are sensitive to the use of preventive care. Although our preventive physician visit variable is not service-specific, our measure of childhood growth and development checks should correspond directly with infectious disease prevalence. Columns three through six of Table 4 show estimates for child health outcomes linked to preventive care use. SR enrollment

³⁴ This increase in preventive care use dominates any *ex ante* moral hazard (which would reduce prevention – although Section IVE suggests no evidence of *ex ante* moral hazard).

³⁵ See Fogel (1994), Santos et al. (2001), Alderman (2007), Currie, Decker, and Lin (2008), Galasso and Umapathi (2009), and Linnemayr and Alderman (2011) on the relationship between preventive care and child health.

is associated with 1.4 fewer child days absent from usual activities due to illness in the past month ($\alpha=0.05$). Enrollment is also associated with an 18 percentage point reduction in the self-reported incidence of cough, fever, or diarrhea among children in the preceding two weeks (but it is not statistically significant at conventional levels). Appendix D Figure 2 shows graphical versions of the intent-to-treat analyses for these health outcomes as well.³⁶ Because these health gains can be linked to increases in preventive service use, they are also likely to improve welfare.

D. Use of Curative Medical Care

As Section II discusses, there are a variety of competing incentives and other forces influencing curative medical care under the SR. These include reductions in out-of-pocket prices, increases in the use of preventive services, and supply-side incentives for limiting total medical spending. Predictions about changes in curative service use under the SR – and their welfare implications – are therefore ambiguous, but they remain a central interest of policymakers and are important for assessing the costs of Colombia’s innovative health insurance reform.

Table 5 reports estimates for various categories of curative care obtained from equations (3) and (4) (Appendix D Figure 3 shows graphical versions of the intent-to-treat analyses). We find that SR enrollment is associated with a 13 percentage point increase in reported physician visits because of health problems within the past 30 days ($\alpha=0.05$). There is no change in use of curative care

³⁶ Interestingly, for some of the more statistically pronounced results (such as physician visits for preventive care and children’s healthy days lost due to illness), the graphical pattern of results shown in Figure 2 Appendix D closely resemble the graphical pattern enrollment in the SR shown in Figure 1: a more or less flat relationship to the left of the threshold, and a monotonic relationship to the right of the threshold. Although suggestive, these shape similarities must still be interpreted cautiously.

among children, which may be due to improvements in child health associated with SR enrollment (as Table 4 shows). Finally, there is no meaningful relationship at conventional significance levels between participation in the SR and hospitalizations.

E. Ex Ante Moral Hazard and Insurance Crowd-Out

Protection from financial risk associated with unexpected illness weakens private incentives for costly health protection (*ex ante* moral hazard) (Pauly 1968). Because we find evidence of greater risk protection, we investigate how protective private health behaviors not directly linked to medical care change with SR enrollment.³⁷ As Table 6 Panel A and Appendix D Figure 4 show, however, we find no meaningful change in breastfeeding or maternal investments in fetal health (alcohol, drug, or tobacco use during pregnancy; or prenatal dietary supplementation with iron, calcium, or folic acid), suggesting little *ex ante* moral hazard associated with SR enrollment.³⁸

Manipulation of official SISBEN scores suggests that Colombians perceive benefits of SR enrollment, so we also investigate the possibility that SR enrollment displaces other forms of explicit health insurance.³⁹ Table 6 Panel B also presents intent-to-treat (ITT) estimates obtained by estimating equation (3) with dichotomous indicators for *Regimen Contributivo* enrollment, other forms of health insurance (those for the military, police officers, and certain industrial

³⁷ More generally, private health behaviors and public health services could theoretically be either complements or substitutes for publicly provided health services. While reductions in the price of medical care may raise the return to private health investments given competing risks, cheaper health services could also instead ‘crowd-out’ costly private health behaviors (Dow, Holmes, Philipson, and Sala-i-Martin 1999, Murphy and Topel 2003).

³⁸ We cannot rule-out the possibility that social desirability bias in reporting might attenuate a true effect.

³⁹ Formal sector employees are mandated to enroll in an employment-based health insurance system called *Regimen Contributivo*. This mandate holds even for individuals with SISBEN scores falling below the SISBEN eligibility threshold for the SR.

groups like oil industry workers, for example), and “uninsurance” as dependent variables. The *Regimen Contributivo* estimates are generally small (-0.02 in the ECV and -0.05 in the DHS) and the latter is statistically significant ($\alpha=0.05$), suggesting that some modest crowding-out of formal insurance may have occurred.

F. Robustness

To investigate the robustness of our results, we estimate a variety of alternative parametric and non-parametric specifications based on our main estimating equations. First, we re-estimate equations (3) and (4) using SISBEN score bandwidths ranging between two and four. Second, at each bandwidth we estimate specifications that include higher order polynomials of relative SISBEN scores (including squared, cubic, and fourth power terms) as well as models that include interactions between relative SISBEN scores and simulated eligibility (although allowing SISBEN gradients to vary on either side of the eligibility threshold should matter little given our narrow bandwidth). Third, we re-estimate specifications that do not include county fixed effects (at each bandwidth), allowing us also to make cross-county comparisons among individuals with identical simulated SISBEN scores but that fall on opposite sides of county-specific eligibility thresholds. Finally, we also estimate models using non-parametric local linear regression.⁴⁰ As Tables 1-4 of Appendix E show that our results are generally robust (with some variation in precision) across these alternative bandwidths and specifications.

⁴⁰ Specifically, we use local linear regression functions with triangle kernels to estimate conditional means of outcome variables (conditioning on *SISBEN_diff*) on either side of the eligibility threshold, and we repeat this estimation process for enrollment in the SR. We then construct Wald statistics using differences in the estimated conditional means of the outcome variable on either side of the threshold as numerators and the difference in the conditional mean of SR enrollment on either side of the threshold as the denominator. We estimate our standard errors using 250 bootstrap replications. Details of the implementation can be found in Nichols (2007).

G. Heterogeneity and External Validity

While RD estimates can have good internal validity, their external validity is often limited because they typically utilize a sample only in the neighborhood of the discontinuity (Imbens and Lemieux 2008, Lee and Lemieux 2010). However, because SR eligibility thresholds vary by county, our empirical strategy allows us to investigate the external validity of our estimates by extending equation (4) to include interactions between absolute SISBEN score and SR enrollment.⁴¹ This allows us to test for heterogeneity by SISBEN score, a summary measure of socio-economic status. As Appendix F Tables 1-4 show, we find little evidence of heterogeneous effects, suggesting that our estimates are not sensitive to local variation in SISBEN scores and may therefore apply to a broader range of the Colombian population.⁴² We note, however, that the interaction terms' standard errors are large, so our power to detect heterogeneity is limited.

⁴¹ Specifically, we estimate: $outcome_{ihc} = \varphi + \lambda enroll_{ihc} + \theta SISBEN_h + \delta(enroll_{ihc} \times SISBEN_h) + \psi SISBEN_diff_{hc} + \sum_k \pi_k estrato_{hk} + \mu_c + \xi_{ihc}$, using $below_{hc}$ and $(below_{hc} \times SISBEN_h)$ as instruments. All variables are defined as in equations (3) and (4).

⁴² The fifth row of Appendix F Tables 1-4 shows implied effects of Subsidized Regime enrollment evaluated at the average SISBEN score (45) and their accompanying p-values. The magnitude and significance of these effect sizes generally match the estimates in Tables 3-6.

V. Conclusion

This paper studies the role of Colombia's *Régimen Subsidiado* in protecting the poor against financial risk, influencing their use of health services, and ultimately affecting their health status. Unusual among middle- and low-income countries at the time of its introduction, the SR is a publicly-financed health insurance program with features of managed care emphasizing allocative efficiency that are traditionally only found in wealthy countries. We find evidence that the SR provides some protection against financial risk and is associated with greater use of preventive services having positive externalities – as well as with measurable health gains among children.

Pinpointing the mechanisms through which complex insurance programs such as SR work is an important topic of future research. In our case, because preventive services are generally free regardless of insurance status, one plausible interpretation of the increase in prevention that we observe links them to the SR's high-powered supply-side incentives (if providers with capitated contracts believe that prevention is cost-saving). Another potentially important channel is patient substitution from public to private sector providers contracting with insurance plans under the SR (if private providers were more likely to supply preventive care). Although we suspect that these two mechanisms may be most salient, we cannot rule out others (individual feeling more comfortable going to the doctor when insured, the provision of insurance making the universal entitlement to preventive care more salient, etc.).

We conclude by emphasizing that the welfare-improving potential of SR has yet to be fully realized. For example, a variety of political concessions followed the creation of the SR – including exemptions from the end of government subsidies as well as requirements that insurers contract with public facilities for a minimum

share of the services that they finance. These concessions presumably limit the ability of health plans to contract with medical care providers in ways that best encourage higher quality and lower cost services.

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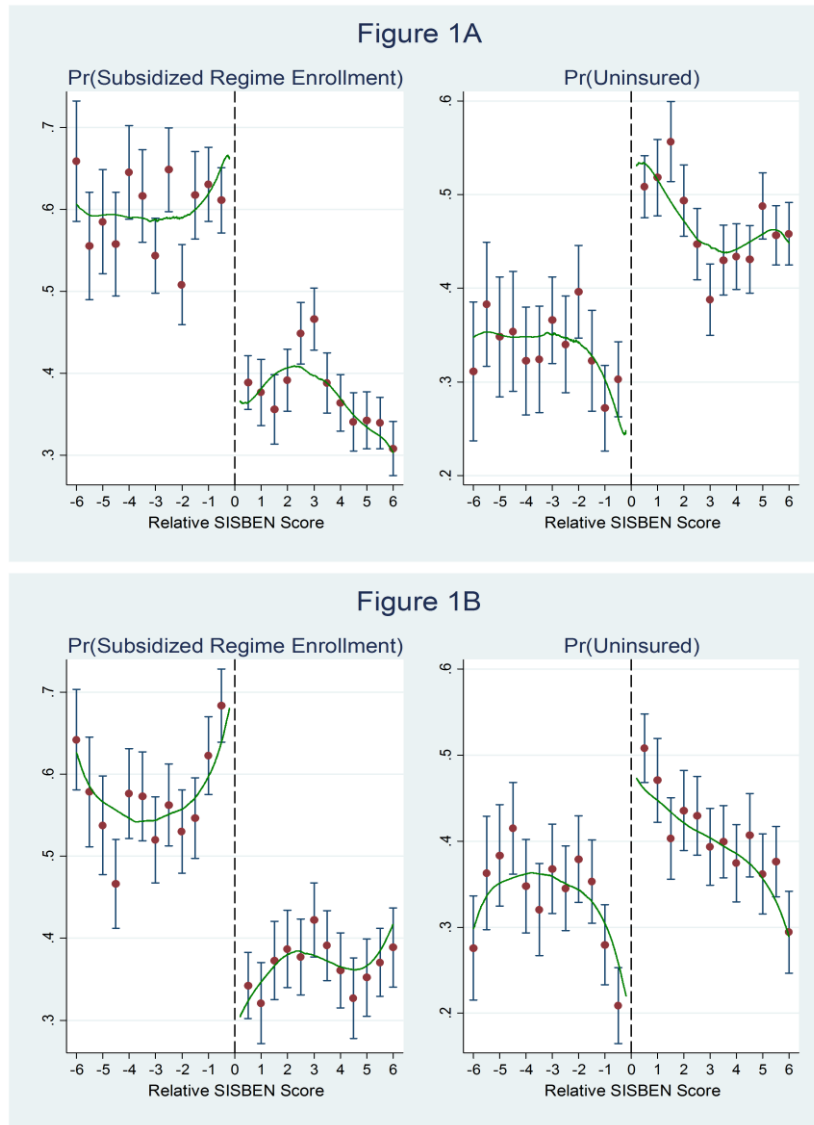
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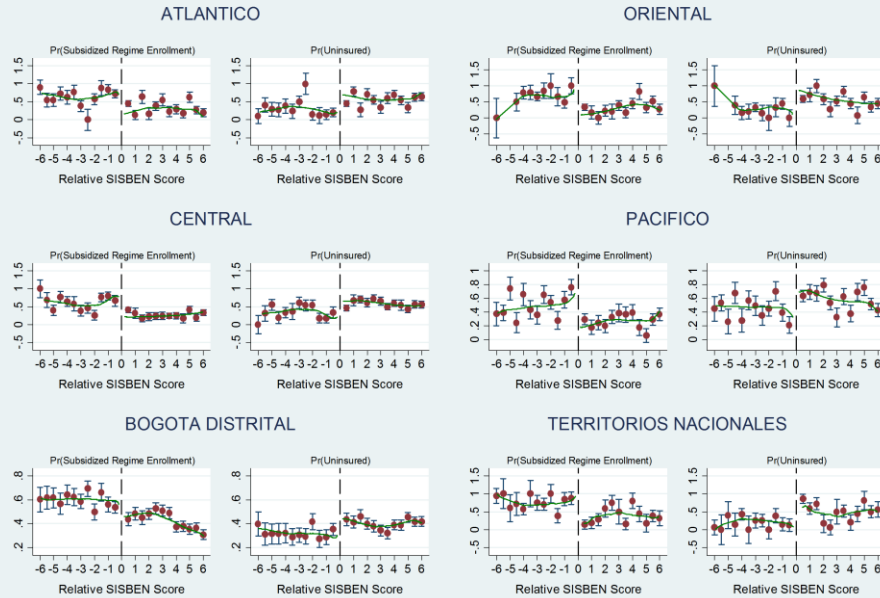
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**FIGURE 1. SUBSIDIZED REGIME ENROLLMENT AND 'UNINSURANCE'
ACROSS SISBEN ELIGIBILITY THRESHOLDS**

Notes: Source of Figure A is ECV 2003 and of Figure B is DHS 2005

Regions - ECV 2003



Regions - DHS WOMEN 2005

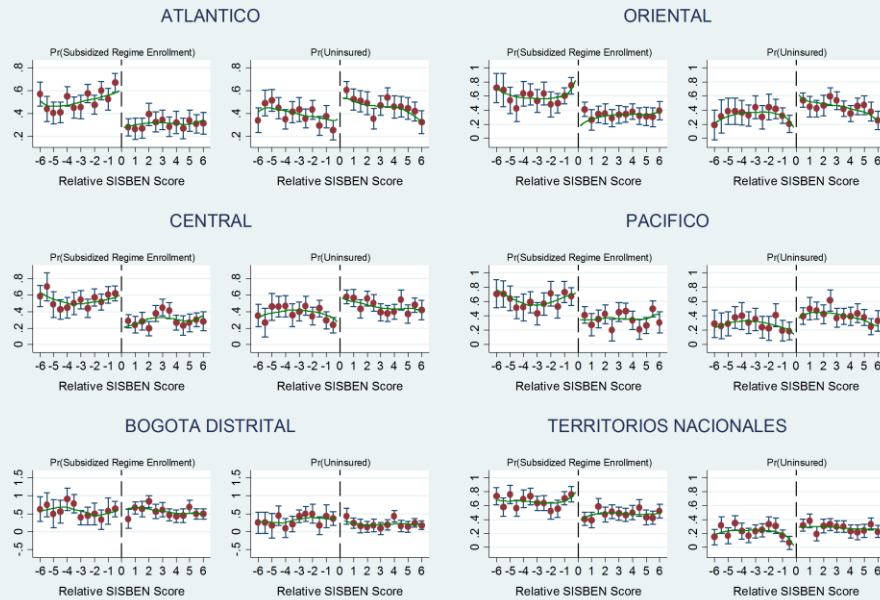


FIGURE 2. SUBSIDIZED REGIME ENROLLMENT AND 'UNINSURANCE' ACROSS SISBEN ELIGIBILITY THRESHOLDS BY REGION

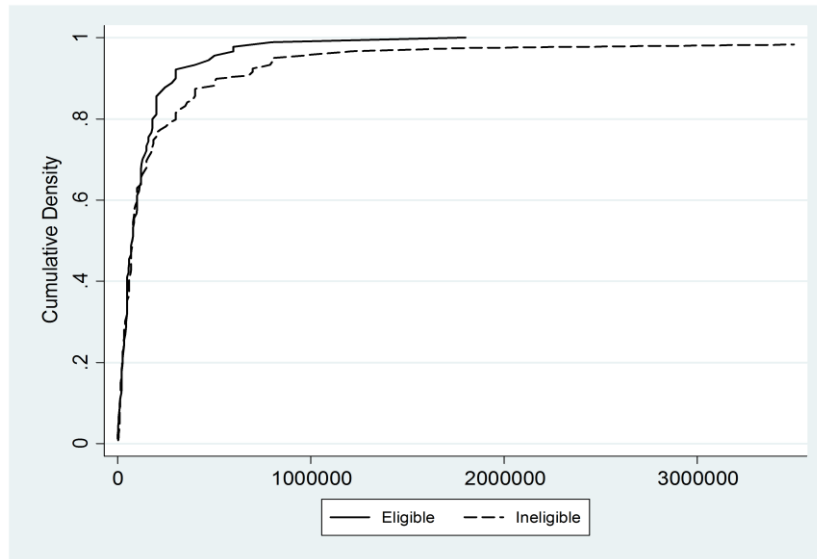


FIGURE 3. INDIVIDUAL INPATIENT MEDICAL SPENDING (AMONG THOSE WITHIN TWO POINTS OF COUNTY-SPECIFIC ELIGIBILITY THRESHOLDS)

Notes: Source ECV 2003

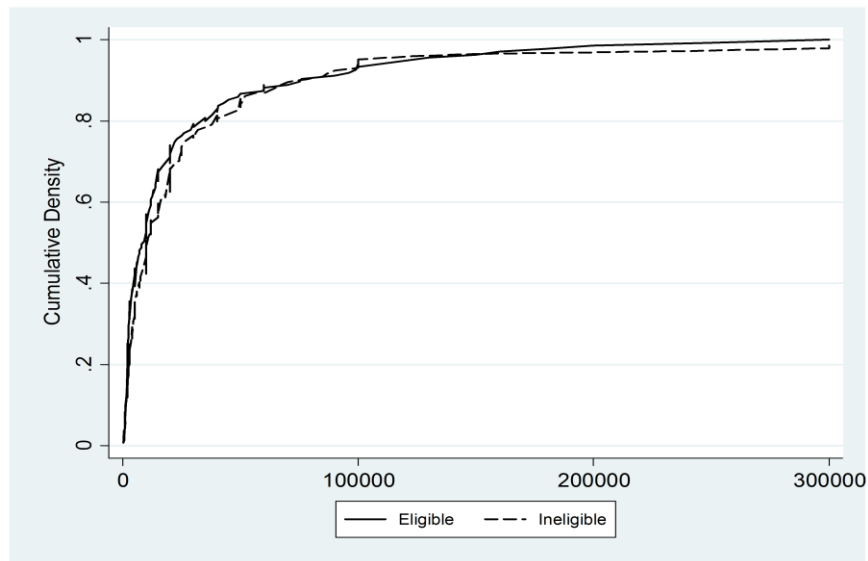


FIGURE 4. INDIVIDUAL OUTPATIENT MEDICAL SPENDING (AMONG THOSE WITHIN TWO POINTS OF COUNTY-SPECIFIC ELIGIBILITY THRESHOLDS)

Notes: Source ECV 2003

TABLE 1—DESCRIPTIVE STATISTICS

Variable	Total			Data
	Mean	SD	N	Source
<i>Outcome Variables</i>				
<u>Risk Protection, Consumption Smoothing, and Portfolio Choice</u>				
Individual Inpatient Medical Spending	10,794	129,777	4219	ECV
Individual Outpatient Medical Spending	2,118	22,565	4218	ECV
Variability of Individual Inpatient Medical Spending	21,935	127,975	4219	ECV
Variability of Individual Outpatient Medical Spending	5,477	21,993	4218	ECV
Individual Education Spending	7,588	20,279	3567	ECV
Household Education Spending	35,145	48,468	4222	ECV
Total Spending on Food	274,881	217,392	4096	ECV
Total Monthly Expenditure	654,552	599,861	966	ECV
Has Car	0.03	0.16	3334	DHS
Has Radio	0.61	0.49	3334	DHS
<u>Medical Care Use</u>				
Preventive Physician Visit	0.47	0.50	4222	ECV
Number of Growth Development Checks Last Year	1.16	1.81	1167	DHS
Curative Care Use (Not Conditional on Health Status)	0.06	0.24	4222	ECV
Primary Care	0.05	0.21	4222	ECV
Medical Visit - Specialist	0.02	0.12	4222	ECV
Hospital Stay	0.08	0.26	4222	ECV
Medical Visit for Chronic Disease	0.58	0.49	564	ECV
Curative Care Use among Children (Not Conditional on Health Status)	0.31	0.46	1161	DHS
<u>Health Status</u>				
Child Days Lost to Illness	0.60	2.01	1161	DHS
Cough, Fever Diarrhea	0.56	0.50	1167	DHS
Any Health Problem	0.65	0.48	1161	DHS
Birthweight (KG)	3.26	0.57	897	DHS
<u>Behavioral Distortions</u>				
Drank Alcohol during Pregnancy	0.11	0.31	998	DHS
Number of Drinks per Week during Pregnancy	2.75	9.76	110	DHS
Months Child Breastfed	12.58	10.87	946	DHS
Folic Acid During Pregnancy	0.54	0.50	988	DHS
Number Months Folic Acid during Pregnancy	3.96	2.42	515	DHS
Contributory Regime Enrollment	0.09	0.29	4222	ECV
Contributory Regime Enrollment	0.14	0.34	3334	DHS
Other Health Insurance	0.44	0.50	4222	ECV
Other Health Insurance	0.01	0.09	3334	DHS
Uninsured	0.00	0.06	4222	ECV
Uninsured	0.38	0.49	3334	DHS
<i>Baseline Demographic Variables</i>				
<u>Balance</u>				
Household Head Age	46.81	14.81	3334	DHS
Household Head Age	46.58	14.79	4222	ECV
Completed Elementary School	0.18	0.39	3333	DHS
Completed Elementary School	0.19	0.39	3764	ECV
Completed Secondary School	0.20	0.40	3333	DHS
Completed Secondary School	0.07	0.26	3764	ECV
Household Head Completed Elementary School	0.30	0.46	3334	DHS
Household Head Completed Elementary School	0.28	0.45	4222	ECV
Household Head Completed Secondary School	0.02	0.15	3334	DHS
Household Head Completed Secondary School	0.02	0.13	4222	ECV
Student Received School Grant	0.08	0.26	1305	ECV
Benefits to Buy House	0.01	0.08	4222	ECV
Attended Training	0.05	0.21	3010	ECV
Household in Hogar Comunitario program	0.10	0.30	4222	ECV
Services from Bienstar Familiar	0.18	0.38	4222	ECV

Notes: All data summarized is from samples of "urban" individuals within two SISBEN index points of county-specific eligibility thresholds in either the 2003 Encuesta de Calidad de Vida (ECV) or the 2005 Demographic and Health Survey (DHS) (as indicated in the last column). SISBEN index calculations are described in detail in Appendix 2, and estimation of county-specific eligibility thresholds is presented in Section IIIB. The first group of three columns present summary statistics for the entire bandwidth of 2 sample, the second group for those not enrolled in the Subsidized Regime, and the third group for those enrolled in the Subsidized Regime.

TABLE 1 (CONT.)—DESCRIPTIVE STATISTICS

Variable	Not Enrolled in the Subsidized Regime			Enrolled in the Subsidized Regime			Data Source
	Mean	SD	N	Mean	SD	N	
Outcome Variables							
Risk Protection, Consumption Smoothing, and Portfolio Choice							
Individual Inpatient Medical Spending	11,360	129,227	2252	10,145	130,435	1967	ECV
Individual Outpatient Medical Spending	2,513	20,170	2249	1,666	25,019	1969	ECV
Variability of Individual Inpatient Medical Spending	26,338	126,615	2252	16,894	129,363	1967	ECV
Variability of Individual Outpatient Medical Spending	7,343	19,017	2249	3,347	24,796	1969	ECV
Individual Education Spending	7,501	23,234	1874	7,684	16,407	1693	ECV
Household Education Spending	34,089	52,464	2253	36,352	43,428	1969	ECV
Total Spending on Food	279,128	201,735	2171	270,091	233,756	1925	ECV
Total Monthly Expenditure	688,065	716,382	479	621,590	455,559	487	ECV
Has Car	0.03	0.18	1758	0.02	0.15	1576	DHS
Has Radio	0.60	0.49	1758	0.62	0.49	1576	DHS
Medical Care Use							
Preventive Physician Visit	0.39	0.49	2253	0.57	0.49	1969	ECV
Number of Growth Development Checks Last Year	1.00	1.75	683	1.38	1.87	484	DHS
Curative Care Use (Not Conditional on Health Status)	0.06	0.23	2253	0.07	0.26	1969	ECV
Primary Care	0.04	0.20	2253	0.06	0.23	1969	ECV
Medical Visit - Specialist	0.01	0.12	2253	0.02	0.13	1969	ECV
Hospital Stay	0.07	0.25	2253	0.08	0.28	1969	ECV
Medical Visit for Chronic Disease	0.52	0.50	262	0.64	0.48	302	ECV
Curative Care Use among Children (Not Conditional on Health Status)	0.29	0.45	680	0.34	0.48	481	DHS
Health Status							
Child Days Lost to Illness	0.65	2.14	680	0.53	1.82	481	DHS
Cough, Fever Diarrhea	0.56	0.50	682	0.56	0.50	485	DHS
Any Health Problem	0.64	0.48	680	0.66	0.47	481	DHS
Birthweight (KG)	3.25	0.53	534	3.27	0.63	363	DHS
Behavioral Distortions							
Drank Alcohol during Pregnancy	0.12	0.32	504	0.10	0.30	494	DHS
Number of Drinks per Week during Pregnancy	3.15	10.85	60	2.28	8.36	50	DHS
Months Child Breastfed	11.15	10.18	578	14.83	11.53	368	DHS
Folic Acid During Pregnancy	0.56	0.50	500	0.53	0.50	488	DHS
Number Months Folic Acid during Pregnancy	4.05	2.39	266	3.86	2.45	249	DHS
Contributory Regime Enrollment	0.17	0.38	2253	0.00	0.00	1969	ECV
Contributory Regime Enrollment	0.26	0.44	1758	0.00	0.00	1576	DHS
Other Health Insurance	0.82	0.39	2253	0.00	0.00	1969	ECV
Other Health Insurance	0.01	0.12	1758	0.00	0.00	1576	DHS
Uninsured	0.01	0.08	2253	0.00	0.00	1969	ECV
Uninsured	0.73	0.45	1758	0.00	0.00	1576	DHS
Baseline Demographic Variables							
Balance							
Household Head Age	47.36	15.37	1758	46.18	14.14	1576	DHS
Household Head Age	45.71	15.12	2253	47.57	14.35	1969	ECV
Completed Elementary School	0.18	0.38	1757	0.19	0.39	1576	DHS
Completed Elementary School	0.18	0.39	1985	0.19	0.39	1779	ECV
Completed Secondary School	0.21	0.41	1757	0.19	0.39	1576	DHS
Completed Secondary School	0.08	0.27	1985	0.07	0.25	1779	ECV
Household Head Completed Elementary School	0.29	0.45	1758	0.31	0.46	1576	DHS
Household Head Completed Elementary School	0.27	0.45	2253	0.28	0.45	1969	ECV
Household Head Completed Secondary School	0.02	0.13	1758	0.03	0.18	1576	DHS
Household Head Completed Secondary School	0.01	0.11	2253	0.02	0.15	1969	ECV
Student Received School Grant	0.05	0.22	651	0.10	0.30	654	ECV
Benefits to Buy House	0.00	0.06	2253	0.01	0.10	1969	ECV
Attended Training	0.06	0.23	1593	0.04	0.19	1417	ECV
Household in Hogar Comunitario program	0.09	0.29	2253	0.11	0.31	1969	ECV
Services from Bienstar Familiar	0.16	0.37	2253	0.20	0.40	1969	ECV

Notes: All data summarized is from samples of "urban" individuals within two SISBEN index points of county-specific eligibility thresholds in either the 2003 Encuesta de Calidad de Vida (ECV) or the 2005 Demographic and Health Survey (DHS) (as indicated in the last column). SISBEN index calculations are described in detail in Appendix 2, and estimation of county-specific eligibility thresholds is presented in Section IIIB. The first group of three columns present summary statistics for the entire bandwidth of 2 sample, the second group for those not enrolled in the Subsidized Regime, and the third group for those enrolled in the Subsidized Regime.

TABLE 2—BALANCE ACROSS ELIGIBILITY THRESHOLDS

Panel A								
Outcome	Household Head Age	Household Head Age	Completed Elementary School	Completed Elementary School	Completed Secondary School	Completed Secondary School	Household Head Completed Elementary School	Household Head Completed Elementary School
2SLS Estimate, Subsidized Regime Enrollment	4.38 [3.05]	3.05 [7.68]	0.01 [0.08]	-0.15 [0.10]	0.03 [0.07]	0.05 [0.05]	-0.09 [0.12]	-0.37 [0.24]
Intent to Treat Estimate	1.77 [1.22]	0.79 [1.84]	0.01 [0.03]	-0.04** [0.02]	0.01 [0.03]	0.01 [0.01]	-0.04 [0.05]	-0.10** [0.05]
First Stage Estimate, Below Eligibility Threshold	0.40*** [0.04]	0.26*** [0.07]	0.40*** [0.04]	0.25*** [0.06]	0.40*** [0.04]	0.25*** [0.06]	0.40*** [0.04]	0.26*** [0.07]
First Stage F-Statistic	125.75	14.08	125.91	17.02	125.91	17.02	125.75	14.08
OLS Estimate	-0.04 [0.66]	1.62*** [0.55]	0.00 [0.01]	0.00 [0.01]	-0.01 [0.02]	-0.01 [0.01]	0.02 [0.02]	-0.01 [0.02]
Mean for those Not Enrolled in the Subsidized Regime	47.36	45.71	0.18	0.18	0.21	0.08	0.29	0.27
Observations	3,334	4,222	3,333	3,764	3,333	3,764	3,334	4,222
Data Source	DHS	ECV	DHS	ECV	DHS	ECV	DHS	ECV
Panel B								
Outcome	Household Head Completed Secondary School	Household Head Completed Secondary School	Student Received School Grant	Benefits to Buy House	Attended Training	Household in Hogar Comunitario Program	Services from Bienestar Familiar	
2SLS Estimate, Subsidized Regime Enrollment	-0.02 [0.03]	-0.04 [0.07]	-0.06 [0.11]	0.02 [0.01]	0.01 [0.04]	0.03 [0.08]	-0.04 [0.13]	
Intent to Treat Estimate	-0.01 [0.01]	-0.01 [0.02]	-0.01 [0.02]	0.00** [0.00]	0.00 [0.01]	0.01 [0.02]	-0.01 [0.03]	
First Stage Estimate, Below Eligibility Threshold	0.40*** [0.04]	0.26*** [0.07]	0.21*** [0.09]	0.26*** [0.07]	0.27*** [0.06]	0.26*** [0.07]	0.26*** [0.07]	
First Stage F-Statistic	125.75	14.08	5.17	14.08	23.45	14.08	14.08	
OLS Estimate	0.01* [0.01]	0.01 [0.01]	0.03 [0.02]	0.00** [0.00]	-0.02*** [0.00]	0.01 [0.01]	0.02 [0.02]	
Mean for those Not Enrolled in the Subsidized Regime	0.02	0.01	0.05	0.00	0.06	0.09	0.16	
Observations	3,334	4,222	1,305	4,222	3,010	4,222	4,222	
Data Source	DHS	ECV	ECV	ECV	ECV	ECV	ECV	

Notes: Individual-level “urban” data for those within two SISBEN index points of county-specific eligibility thresholds from the 2003 ECV and 2005 DHS. Dependent variables are shown at the top of each column. The first row shows 2SLS estimates for enrollment in the Subsidized Regime (SR), instrumenting for SR enrollment using simulated eligibility. The second row shows reduced-form intent-to-treat estimates for simulated SISBEN score falling below the county-specific eligibility threshold. The third row shows first stage estimates from OLS regressions of Subsidized Regime enrollment on an indicator for falling below the county-specific eligibility threshold. All specifications

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

* Significant at the 10 percent level.

Table 3—RISK PROTECTION AND PORTFOLIO CHOICE

Panel A: Risk Protection						
Outcome	Individual Inpatient Medical Spending	Individual Outpatient Medical Spending	Variability of Individual Inpatient Medical Spending	Variability of Individual Outpatient Medical Spending		
2SLS Estimate, Subsidized Regime Enrollment	-58,870* [33,263]	3,562 [2,702]	-67,499.38** [32,906]	167.57 [2,417]		
Intent to Treat Estimate	-15,108* [8,888]	918.23 [821]	-17322.90* [9,120]	43.20 [626]		
First Stage Estimate, Below Eligibility Threshold	0.26*** [0.07]	0.26*** [0.07]	0.26*** [0.07]	0.26*** [0.07]		
First Stage F-Statistic	13.91	14.01	13.91	14.01		
OLS Estimate	-5,655 [3,898]	-1,204*** [342]	-13,888*** [3,893]	-4,387*** [357]		
Mean for those Not Enrolled in the Subsidized Regime	11,359.86	2,512.98	26338.40	7342.59		
Observations	4,219	4,218	4,219	4,218		
Data Source	ECV	ECV	ECV	ECV		
Panel B: Portfolio Choice						
Outcome	Individual Education Spending	Household Education Spending	Total Spending on Food	Total Monthly Expenditure	Has Car	Has Radio
2SLS Estimate, Subsidized Regime Enrollment	-341.68 [3,781]	30,366 [25,055]	32,136 [103,540]	-33,826 [278,060]	0.01 [0.04]	0.17 [0.11]
Intent to Treat Estimate	-84.72 [945]	7,815 [4,880]	8,709 [28,491]	-14,036 [115,736]	0.01 [0.01]	0.07 [0.05]
First Stage Estimate, Below Eligibility Threshold	0.25*** [0.06]	0.26*** [0.07]	0.27*** [0.06]	0.41*** [0.12]	0.40*** [0.04]	0.40*** [0.04]
First Stage F-Statistic	19.28	14.08	18.80	12.18	125.75	125.75
OLS Estimate	122.82 [231]	2,952.32*** [902]	-12,036 [10,330]	-39,273 [58,730]	-0.01 [0.01]	0.03 [0.02]
Mean for those Not Enrolled in the Subsidized Regime	7,501	34,089	279,128	688,065	0.03	0.60
Observations	3,567	4,222	4,096	966	3,334	3,334
Data Source	ECV	ECV	ECV	ECV	DHS	DHS

Notes: Individual-level “urban” data for those within two SISBEN index points of county-specific eligibility thresholds from the 2003 ECV and 2005 DHS. Dependent variables are shown at the top of each column. The first row shows 2SLS estimates for enrollment in the Subsidized Regime (SR), instrumenting for SR enrollment using simulated eligibility. The second row shows reduced-form intent-to-treat estimates for simulated SISBEN score falling below the county-specific eligibility threshold. The third row shows first stage estimates from OLS regressions of Subsidized Regime enrollment on an indicator for falling below the county-specific eligibility threshold. All specifications also include SISBEN score, distance from the county-specific threshold, estrato dummy variables, and county fixed effects. Standard errors (clustered by county) are shown in brackets below each estimate.

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

* Significant at the 10 percent level.

Table 4—USE OF PREVENTIVE MEDICAL CARE AND HEALTH STATUS

Outcome	Use of Preventive Care		Health Status (Children)			
	Preventive Physician Visit	Number of Growth Dev. Checks Last Year	Child Days Lost to Illness	Cough, Fever, Diarrhea	Any Health Problem	Birthweight (KG)
2SLS Estimate, Subsidized Regime Enrollment	0.29*** [0.11]	1.50** [0.69]	-1.40** [0.65]	-0.18 [0.17]	-0.06 [0.18]	0.26 [0.29]
Intent to Treat Estimate	0.08*** [0.03]	0.55** [0.25]	-0.49** [0.20]	-0.07 [0.06]	-0.02 [0.06]	0.11 [0.12]
First Stage Estimate, Below Eligibility Threshold	0.26*** [0.07]	0.36*** [0.07]	0.35*** [0.07]	0.37*** [0.07]	0.35*** [0.07]	0.41*** [0.09]
First Stage F-Statistic	14.08	25.24	23.46	25.19	23.46	19.10
OLS Estimate	0.17*** [0.01]	0.33*** [0.12]	-0.04 [0.17]	0.01 [0.04]	0.03 [0.04]	0.04 [0.05]
Mean for those Not Enrolled in the Subsidized Regime	0.39	1.00	0.65	0.56	0.64	3.25
Observations	4,222	1,167	1,161	1,167	1,161	897
Data Source	ECV	DHS	DHS	DHS	DHS	DHS

Notes: Individual-level “urban” data for those within two SISBEN index points of county-specific eligibility thresholds from the 2003 ECV and 2005 DHS. Dependent variables are shown at the top of each column. The first row shows 2SLS estimates for enrollment in the Subsidized Regime (SR), instrumenting for SR enrollment using simulated eligibility. The second row shows reduced-form intent-to-treat estimates for simulated SISBEN score falling below the county specific eligibility threshold. The third row shows first stage estimates from OLS regressions of Subsidized Regime enrollment on an indicator for falling below the county-specific eligibility threshold. All specifications also include SISBEN score, distance from the county-specific threshold, estrato dummy variables, and county fixed effects. Standard errors (clustered by county) are shown in brackets below each estimate.

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

* Significant at the 10 percent level.

Table 5—USE OF CURATIVE MEDICAL CARE

Outcome	Curative Use (Not Conditional on Health Status)	Curative Use among Children (Not Conditional on Health Status)	Medical Visit for Chronic Disease	Hospital Stay
2SLS Estimate, Subsidized Regime Enrollment	0.13** [0.05]	-0.00 [0.16]	0.51* [0.28]	-0.04 [0.11]
Intent to Treat Estimate	0.03* [0.02]	0.00 [0.06]	0.18*** [0.06]	-0.01 [0.03]
First Stage Estimate, Below Eligibility Threshold	0.26*** [0.07]	0.35*** [0.07]	0.35*** [0.10]	0.26*** [0.07]
First Stage F-Statistic	14.08	23.46	13.49	14.08
OLS Estimate	0.03*** [0.01]	0.07* [0.04]	0.13*** [0.03]	0.02** [0.01]
Mean for those Not Enrolled in the Subsidized Regime	0.06	0.29	0.52	0.07
Observations	4,222	1,161	564	4,222
Data Source	ECV	DHS	ECV	ECV

Notes: Individual-level “urban” data for those within two SISBEN index points of county-specific eligibility thresholds from the 2003 ECV and 2005 DHS. Dependent variables are shown at the top of each column. The first row shows 2SLS estimates for enrollment in the Subsidized Regime (SR), instrumenting for SR enrollment using simulated eligibility. The second row shows reduced-form intent-to-treat estimates for simulated SISBEN score falling below the county specific eligibility threshold. The third row shows first stage estimates from OLS regressions of Subsidized Regime enrollment on an indicator for falling below the county-specific eligibility threshold. All specifications also include SISBEN score, distance from the county-specific threshold, estrato dummy variables, and county fixed effects. Standard errors (clustered by county) are shown in brackets below each estimate.

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

* Significant at the 10 percent level.

Table 6—BEHAVIORAL DISTORTION – EX ANTE MORAL HAZARD AND INSURANCE CROWD-OUT

Panel A: Ex-Ante Moral Hazard			
Outcome	Drank Alcohol during Pregnancy	Months Breastfed as Child	Folic Acid During Pregnancy
2SLS Estimate, Subsidized Regime Enrollment	-0.10 [0.11]	-0.23 [5.36]	0.10 [0.18]
Intent to Treat Estimate	-0.04 [0.04]	-0.07 [1.64]	0.04 [0.07]
First Stage Estimate, Below Eligibility Threshold	0.40*** [0.07]	0.31*** [0.07]	0.41*** [0.07]
First Stage F-Statistic	28.83	19.00	30.12
OLS Estimate	-0.02 [0.02]	4.02*** [0.91]	0.02 [0.03]
Mean for those Not Enrolled in the Subsidized Regime	0.12	11.15	0.56
Observations	998	946	988
Data Source	DHS	DHS	DHS

Panel B: Insurance Crowd-Out						
Outcome	Contributory Regime Enrollment	Uninsured	Other Health Insurance	Contributory Regime Enrollment	Uninsured	Other Health Insurance
Intent to Treat Estimate	-0.02 [0.02]	-0.23*** [0.06]	-0.00 [0.00]	-0.05** [0.02]	-0.35*** [0.04]	-0.01* [0.01]
Mean for those Not Enrolled in the Subsidized Regime	0.17	0.01	0.82	0.26	0.73	0.01
Observations	4,222	4,222	4,222	3,334	3,334	3,334
Data Source	ECV	ECV	ECV	DHS	DHS	DHS

Notes: Individual-level “urban” data for those within two SISBEN index points of county-specific eligibility thresholds from the 2003 ECV and 2005 DHS. Dependent variables are shown at the top of each column. The first row shows 2SLS estimates for enrollment in the Subsidized Regime (SR), instrumenting for SR enrollment using simulated eligibility. The second row shows reduced-form intent-to-treat estimates for simulated SISBEN score falling below the county-specific eligibility threshold. Intent-to-treat estimates only are reported for the crowd-out analyses of other insurance types (the last six columns). The third row shows first stage estimates from OLS regressions of Subsidized Regime enrollment on an indicator for falling below the county-specific eligibility threshold. All specifications also include SISBEN score, distance from the county-specific threshold, estrato dummy variables, and county fixed effects. Standard errors (clustered by household) are shown in brackets below each estimate.

*** Significant at the 1 percent level.

** Significant at the 5 percent level.

* Significant at the 10 percent level.