

SOCIAL PROTECTION AMID A CRISIS: NEW EVIDENCE FROM SOUTH AFRICA'S OLDER PERSON'S GRANT*

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Abstract

We study the effects of South Africa's Older Person's Grant program on economic and psychological well-being amid the COVID-19 pandemic. With household-level panel data collected before and during the pandemic we leverage the age-eligibility threshold of the grant to implement a local randomization estimation approach. Prior to the pandemic, we find that grant receipt improves household-level economic well-being. During the pandemic, we find that the grant allowed recipient households to better manage the widespread adverse consequences of the pandemic. Specifically, recipient households were less likely to report (i) running out of money for food, (ii) hunger among either adults or children, and (iii) psychological distress. These results provide critical insight into the effectiveness of one of the world's most well-known social protection programs during a massive global health crisis.

Keywords: Cash transfer, Pension, Hunger, Social protection, COVID-19, South Africa, Food expenditures, and Psychological distress

JEL Codes: O21, O38, I15, and I38.

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1 Introduction

The SARS-CoV2 coronavirus (COVID-19) pandemic generates new motivation for understanding the design, reach, and effects of large scale social protection and cash transfer programs. This is especially true in low- and middle-income countries where access to COVID-19 vaccinations remains limited. More than a year and a half into the pandemic, for example, only 1.2 percent of people in Sub-Saharan Africa had been fully vaccinated against the COVID-19 virus (Miguel and Mobarak, 2021). With low vaccination rates and the threat of new variants of the virus, the adverse social, economic, and psychological consequences of the pandemic may extend for years. Therefore, understanding the effectiveness of social protection programs in assisting households mitigate the adverse consequences of the pandemic is exceedingly important for informing effective policy responses.

In this paper we study how a large and wide-reaching cash transfer program allowed recipient households manage the adverse socio-economic consequences of a major health crisis—the COVID-19 pandemic. We specifically show that households with a member receiving South Africa’s Older Person’s grant were able to shield themselves from hunger and psychological distress during the pandemic. The Older Person’s Grant program (also known as the Old Age Pension) is one of the most well-established and well-known social protection programs in the world. It is a means-tested unconditional cash transfer program for the elderly where recipients who are at least 60 years old receive up to 1,800 South African Rand per month—a sum that is nearly 140 percent of the median per capita income in the country and almost double the income poverty line.¹

Using data collected during the pandemic linked to a large ongoing panel study collected prior to the pandemic, we leverage the age-eligibility threshold of the Older Person’s Grant and use a local randomization approach to credibly estimate the effect of grant receipt on the household before and during the pandemic.² Our data come from the National Income Dynamics Study (NIDS) of South Africa—a panel study following households from an initial sample in 2008 approximately every two years through 2017. We also use data from the Coronavirus Rapid Mobile (CRAM) survey, a phone-based survey administered to a random sub-set of individuals from the fifth wave of the NIDS. The CRAM survey was designed to study the consequences of the COVID-19 pandemic and includes five waves administered in 2020 and 2021.

¹This is equivalent to approximately 120 US dollars per month and about 15 percent of average household income per month in South Africa.

²Our econometric approach is akin to a fuzzy regression discontinuity design but with a small number of mass points (Cattaneo, Idrobo and Titiunik, forthcoming).

We first document the impact of the Older Person’s Grant program in pre-pandemic waves of the NIDS data. In doing so, we update various stylized facts about the program originally documented by [Case and Deaton \(1998\)](#) and show that the program remains well-targeted and has wide reach, especially to relatively poor households with children. In addition, a substantial portion of the grant is spent on food, especially among poorer households. We then leverage the discontinuity in grant receipt at age 60 and find that indicators of economic well-being measured at the household level (e.g., household income per capita, household food expenditures per capita, wealth, and hunger) substantially improve with grant receipt.

Our core results examine the effectiveness of the Older Person’s Grant program amid the COVID-19 pandemic. We document three main findings: First, the Older Person’s grant led to a 12.7 percentage point reduction of an individual reporting that their household ran out of money for food in the prior month during the COVID-19 pandemic. Second, receipt of the grant led to a 10 and 7 percentage point reduction in respondents reporting the presence of adult and child hunger within their household respectively—effects that translates to a 40 and 45 percent reduction in reported hunger for adults and children. We find a similar reduction in reported extreme hunger (almost daily) among grant recipients. Finally, CRAM survey respondents with household members who are receiving the grant are less likely to report experiencing psychological distress.

These findings are important for at least two reasons: First, specifically within South Africa, in the initial months of the COVID-19 pandemic the government closed schools and school lunch programs, shutdown informal food vendors, and stretched the food budgets of vulnerable households ([Wills et al., 2020](#); [Arndt et al., 2020](#)). In particular, adult and child hunger were each reported in one out of every three households in our data at the peak of South Africa’s COVID-19 pandemic lockdowns. Second, it is also an important finding more generally as, in response to the pandemic, the number of social protection programs around the world more than doubled between 2020 and 2021, with cash transfers and social pension programs representing over 40 percent of these programs ([Gentilini et al., 2021](#)). To cite just one specific example, policymakers in the United States are aiming to expand child tax credit legislation in response to the COVID-19 pandemic.

Our work is similar in spirit to two recent studies by [Abay et al. \(2021\)](#) and [Londoño-Vélez and Querubin \(2022\)](#), who study the effect of social protection programs amid the COVID-19 pandemic. First, [Abay et al. \(2021\)](#) study the extent to which Ethiopia’s Productive Safety Net program mitigated the adverse consequences of the COVID-19 pandemic on the food security of households. Our work differs in that South Africa’s Older Person’s

Grant program provides passive income to needy households, while Ethiopia’s Productive Safety Net program provides participants with labor-intensive employment opportunities. Second, [Londoño-Vélez and Querubin \(2022\)](#) use a randomized controlled trial to study the effects of a new unconditional cash transfer program implemented by the Colombian government in response to the COVID-19 pandemic. Our work differs in that we study the effects of an existing program amid the pandemic, rather than a new cash transfer program.

We make three main contributions in this paper: First, we contribute to the literature on social protection programs amid the COVID-19 pandemic by specifically investigating the effectiveness of one of the most well-established and well-known social protection programs in the developing world ([Abay et al., 2021](#); [Gentilini et al., 2021](#); [Gulesci, Puente-Beccar and Ubfal, 2021](#)). More generally, these results speak to the role that large, reliable cash transfer programs can play in helping recipient households deal with the effects of large, unexpected shocks ([Londoño-Vélez and Querubin, 2022](#)). Second, we document important socio-economic effects of South Africa’s Older Person’s Grant program that persist despite documented behavior changes associated with grant receipt ([Abel, 2019](#); [Duflo, 2000](#); [Ardington, Case and Hosegood, 2009](#); [Bertrand, Mullainathan and Miller, 2003](#); [Duflo, 2003](#); [Edmonds, Mammen and Miller, 2004](#); [Edmonds, 2006](#); [Hamoudi and Thomas, 2014](#); [Lovo, 2011](#)). We show that despite these documented behavioral changes, several measures of household-level economic well-being improve significantly due to the grant. Finally, we update descriptive findings about the reach of South Africa’s Older Person’s Grant program originally documented by [Case and Deaton \(1998\)](#). We find that the grant continues to predominantly reach poor households and, despite targeting elderly recipients, households where poor children live.

In the next section we briefly introduce the study context by discussing South Africa’s COVID-19 crisis and the Older Person’s Grant program. Section 3, describes the data we use in this paper and our empirical approach. Section 4 investigates the effect of the grant on food expenditure and the role it plays in lowering hunger and mitigating the adverse economic and psychological consequences associated with the COVID-19 pandemic. Finally, section 5 concludes.

2 Study Context

In the immediate aftermath of the onset of the COVID-19 pandemic, rapid analysis using the CRAM survey data revealed the severity of the crisis in South Africa. As reported by

Wills et al. (2020), two out of every five adults responding to the CRAM survey reported that their household had lost its main source of income since the onset of the pandemic; 47 percent of respondents reported running out of money to buy food; 26 percent reported that someone in their household went hungry in the past week; and 15 percent reported that a child in their household went hungry in the past week. This rapid ex-post analysis is qualitatively consistent with the ex-ante analysis reported by Arndt et al. (2020). Furthermore, reporting by *The New York Times* in September 2020 (Goodman, Dahir and Singh, 2020) noted the following about the situation in South Africa:

"When the pandemic emerged in March, the government ordered the shutdown of informal food vendors and township shops, unleashing the military to detain merchants who violated orders. That forced residents to rely on supermarkets — suddenly farther away than ever, given the lockdown of already woeful bus service.

At the same time, South Africa closed its schools, eliminating school lunches — the only reliable meal for millions of students — just as breadwinners lost their means of getting to jobs. By the end of April, nearly half of all South African households had exhausted their funds to buy food, according to an academic study.³ Social unrest eventually prompted a loosening of the country's restrictions."

In response to this crisis, the South African government implemented several expansions to existing social protection programs. As documented by Gentilini et al. (2021), from May through October 2020 South Africa's Child Support Grant expanded by 300 Rand per month, shifted school-feeding programs to take-home food rations, expanded unemployment benefits, and increased wage subsidies. These expansions, in addition to being sorely needed at the time of their implementation, are directly relevant to the results presented here. These social support expansions of other programs will bias any estimated effect of receiving the Older Person's grant toward zero especially if households not receiving the grant are more likely to qualify for other programs. To the extent that these programs also influence household-level indicators of economic well-being and hunger, our results represent estimates of the lower bound of the true effect of the Older Person's Grant program amid the COVID-19 pandemic.

³This study is Wills et al. (2020) who use the CRAM survey access to food and poverty in South Africa amid the early months of the COVID-19 pandemic.

2.1 South Africa's Older Person's Grant

The South Africa Older Person's Grant is South Africa's largest social welfare program and was greatly expanded after the end of Apartheid to target the country's most disadvantaged groups and achieve parity in both eligibility and benefits between whites and Africans (Van der Berg, 1997; Case and Deaton, 1998; Duflo, 2003). It is an unconditional cash-transfer program that every South African citizen or permanent resident becomes eligible for when they turn 60 years old.⁴ While age is the main criteria for eligibility, the program is also means tested based on individual (if single) or combined (if married) income and liquid assets—in practice, income is the main screening criteria.⁵ The transfer amount is approximately ZAR 1,800 a month or nearly 140 percent of the mean household income per capita. Most South Africans (nearly 80 percent) fall below the high means test threshold and take-up rates are high especially among women. Nearly 25 percent of individuals under 60 years of age live with someone who receives this grant, making it an important and far-reaching social safety net in South Africa.

The behavioral effects of the Older Person's Grant have been studied extensively. Seminal work by Case and Deaton (1998) documents several stylized facts about the breadth and depth of the grant program in the early 1990s.⁶ Subsequent research by Duflo (2000) shows improved child health due to the expansion of the Older Person's Grant program to Africans after Apartheid. In a follow-up paper, Duflo (2003) shows that the gender of the recipient matters in determining outcomes for female grandchildren. In two separate studies Bertrand, Mullainathan and Miller (2003) and Abel (2019) show that the labor supply of non-recipient household members within recipient households decreases especially when the recipient is a woman. Somewhat to the contrary, Ardington, Case and Hosegood (2009) show that prime-aged adults increase their labor supply through migration when an elderly household member starts receiving the Older Person's Grant. Additional evidence that the Older Person's Grant helps relieve liquidity constraints are documented by Edmonds (2006), who shows that child labor decreases and school attendance increases among children living with a grant recipient. Moreover, Lovo (2011) shows improvement in technical efficiency among agricultural households receiving the grant. Finally, other

⁴Prior to 2010, females were eligible at age 60 while men became eligible at a later age of 65.

⁵To qualify an individual must (i) be a South African citizen, permanent resident, or refugee, (ii) live in South Africa, (iii) not receive any other social grant, (iv) not be cared for in a state institution, (v) not earn more than 86,280 South African Rand if single or 171,560 South African Rand if married, and (vi) not have assets worth more than 1,227,600 South African Rand if single or 2,455,200 South African Rand if married. Eligibility is not dependent on labor force status.

⁶We update these stylized facts in Section 4.1 of this paper.

studies show household composition changes in response to grant receipt (Hamoudi and Thomas, 2014; Edmonds, Mammen and Miller, 2004).

3 Data and Empirical Framework

We use panel data from the National Income Dynamics Study (NIDS) of South Africa.⁷ The first survey wave of this study was conducted in 2008 and households (and individuals) were interviewed again in 2010, 2012, 2014, and 2017. The 2008 sample of nearly 27,000 individuals is nationally representative.⁸ The NIDS collects data on many socio-economic variables including demographic information, income, consumer expenditure, labor market participation, information on self-employment and farming activity, fertility, health, migration, education, and anthropometric measures.⁹ We specifically use the detailed information on household income, assets, and food expenditures. We mainly choose to use household food expenditure per capita throughout the analysis, but we also use total household income and a wealth index as other measures of economic well-being.

In early 2020, the South Africa Labor and Social Development unit (the group behind the administration of the NIDS) developed the Coronavirus Rapid Mobile (CRAM) survey. The CRAM Survey is a follow-up phone survey of over 7,000 individuals selected from the 2017 wave of the NIDS. The CRAM survey asks a range of questions relating to income, employment, household welfare (i.e., hunger), psychological distress, receipt of grants and social support, and knowledge and behavior relating to the COVID-19 pandemic. We use data from five waves of the CRAM survey, which provide insight into the experience of South African households amid the COVID-19 pandemic.¹⁰

Finally, as a supplementary source of pre-pandemic data, we also use information from the 2016 Demographic and Health Survey (DHS) from South Africa.¹¹ The DHS data provide rich information on a host of demographic and health related topics. We specifically use information on experienced hunger at the household level to supplement the NIDS data and to compare results from the CRAM survey to pre-pandemic, baseline levels of

⁷This is a panel study conducted by the South Africa Labor and Development Research Unit at the University of Cape Town.

⁸15,630 completing the adult individual questionnaire in 6,598 households. Each wave's sample is refreshed in order to deal with attrition and keep each wave nationally representative.

⁹The NIDS survey data are publicly available online: <http://www.nids.uct.ac.za/>.

¹⁰The CRAM survey data are available online: <https://cramsurvey.org/about/>.

¹¹The 2016 DHS sample uses the 2011 South African Census as a sampling frame with enumeration areas from the Census serving as primary sampling units for the DHS sample. The DHS sample uses a two-stage sampling framework that first selects 750 primary sampling units and next randomly selects dwelling units (i.e., households) within primary sampling units.

hunger—however, it is important to note that the samples, while nationally representative, are different and the questions on hunger in the DHS and CRAM survey data are not exactly the same.

3.1 Identification Strategy

Due to endogeneity in the receipt of the Older Person’s Grant, several studies of the Older Person’s Grant limit their sample to a relatively narrow age range around the grant’s age-eligibility threshold (Edmonds, 2006; Ardington, Case and Hosegood, 2009). We follow this approach and also employ an instrumental variable estimation design based on the age-eligibility threshold of the Older Person’s Grant program. This estimation approach requires two conditions: a verifiable data requirement (i.e., the instrument must be relevant) and an assumption (i.e., the instrument must be excludable). The first condition requires that the probability of grant receipt must increase due to eligibility. Figure 1 shows that, at the individual level, there is a large jump in receipt of the grant at age 60, clearly highlighting the relevance of the instrument.¹² The second condition assumes that being eligible for the grant or having another eligible member in one’s household should only affect our dependent variables of interest through the receipt of the grant (i.e., the instrument must be excludable). For the overall NIDS sample, this second condition is not plausible. Having a household member who is 60 years old or older likely changes household preferences and dynamics in many ways that can also affect consumption decisions. We instead rely on the more narrow assumption that having a 59 year-old household member is similar to having a 61 year-old household member—the sole difference being that the member over 60 is eligible for and likely receiving the Older Person’s Grant. Our main variables of interest are food expenditure per capita and hunger and we assume that preferences for food do not change at age 60.¹³

Our identification strategy is based on restricting our sample to households with members who are around the age of 60. This increases the likelihood that we satisfy our second assumption: that being 60 or older or having another household member who is age 60 or older only affects outcome variables of interest through the channel of grant receipt. We show results for samples that are restricted to five different age ranges, all centered on the

¹²In order to identify the sample that would be eligible based on the means test, we use income information to exclude those with reported incomes that would make them ineligible. Through this, we exclude approximately 10% of our sample—among this excluded group, only 5% of individuals above 60 receive the grant whereas among those we keep in our sample more than 94% of individuals above 60 receive the grant.

¹³Figure A.1 in the Appendix shows that the share of income spent on food does not change abruptly when the household head turns 60.

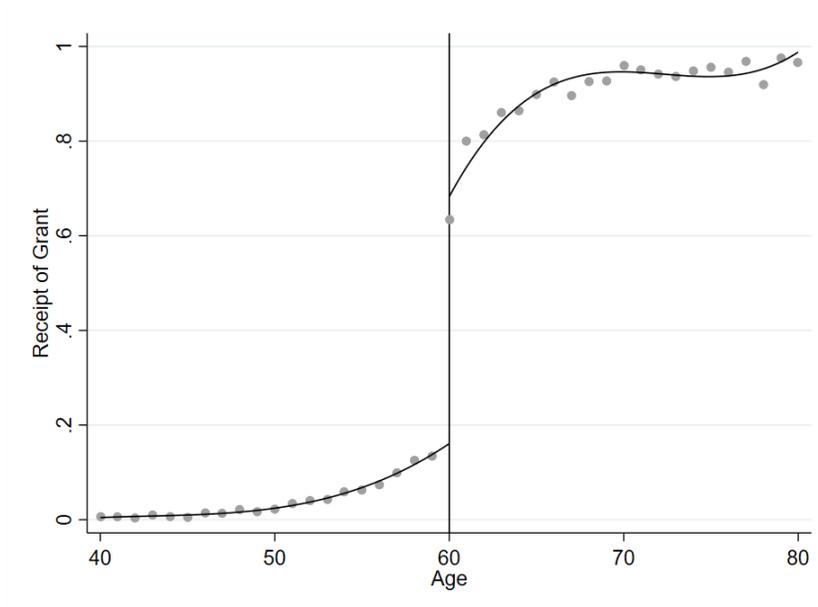


FIGURE 1: Individual-level receipt of the Older Person’s Grant by age. There is a clear discontinuity of grant receipt around the age of eligibility of 60.

age of 60. At its widest, we will use a distance of five where we restrict the sample to individuals in households with a member between the age of 55 and 64 (inclusive).¹⁴ The smallest range of ages is one, where we only keep individuals who are in households with a member who is either 59 or 60 years old. This local randomization approach is akin to a fuzzy regression discontinuity design and is discussed by [Cattaneo, Idrobo and Titiunik \(forthcoming\)](#).¹⁵

Table 1 shows balance tests for the restricted sample that supports the assumption that these households, and the individuals in them, are similar except for eligibility of a member (or members) for receiving the grant. Using data from the NIDS sample, Panel A in Table 1 shows strong balance at the household-level across these two groups: we cannot statistically differentiate the two groups with respect to household size, number of chil-

¹⁴This gives us a bandwidth or window—as is referred to in local randomization approaches in the regression discontinuity literature—of 5 around the age-eligibility cutoff: ages 55, 56, 57, 58, and 59 are in but not eligible for the grant, while 60, 61, 62, 63, and 64 are. Similarly for smaller windows, we successively remove one year from each end.

¹⁵See discussion in [Cattaneo, Idrobo and Titiunik \(forthcoming\)](#): with a small number of mass points around the cutoff, continuity-based regression discontinuity analysis is useful only as an exploratory device without strong parametric assumptions because extrapolation between the mass points becomes unavoidable. In practical terms, the sample size in continuity-based approaches is essentially the number of mass points, which in our case is very small. [Cattaneo, Idrobo and Titiunik \(forthcoming\)](#) suggest local randomization approaches as more appropriate for this type of data.

TABLE 1: Balance Table: Age Range 57-62

	Grant-Eligible Group		Non-Eligible Group		p-value of Δ
	Mean	SE	Mean	SE	
<i>Panel A: NIDS Data</i>					
Household Level					
<i>Household Size</i>	5.43	0.07	5.37	0.07	0.56
<i>Average Age</i>	35.60	0.28	33.21	0.28	0.00
<i>Number of Children</i>	1.86	0.04	1.82	0.04	0.46
<i>Number of Elderly (66+)</i>	0.25	0.01	0.21	0.01	0.00
<i>Urban</i>	0.43	0.01	0.42	0.01	0.59
<i>Death in the past 2 years</i>	0.11	0.01	0.11	0.01	0.96
<i>Total non-grant income per capita (ZAR)</i>	1,087	309	1,098	264	0.79
Variables expected to change					
<i>Old-Age Grant income per capita (ZAR)</i>	376.4	11.94	108.1	7.93	0.00
<i>Savings</i>	0.36	0.01	0.33	0.01	0.05
<i>Share poor</i>	0.32	0.01	0.38	0.01	0.00
Individual Level (Members around threshold)					
<i>Age</i>	61.02	0.04	58.55	0.03	0.00
<i>Male</i>	0.24	0.02	0.25	0.01	0.62
<i>Married</i>	0.44	0.02	0.43	0.02	0.70
<i>In the Labor force</i>	0.00	0.00	0.00	0.00	
<i>Secondary-Level Education</i>	0.10	0.01	0.10	0.01	0.67
<i>Health Issue in the last 30 days</i>	0.70	0.02	0.73	0.01	0.20
Other Adult Household Members (Excluding members around threshold)					
<i>Age</i>	33.65	0.22	33.07	0.25	0.08
<i>Male</i>	0.48	0.01	0.48	0.01	0.56
<i>Married</i>	0.15	0.01	0.15	0.01	0.86
<i>In the Labor force</i>	0.49	0.01	0.47	0.01	0.17
<i>Secondary-Level Education</i>	0.51	0.01	0.52	0.01	0.70
<i>Health Issue in the last 30 days</i>	0.42	0.01	0.40	0.01	0.17
<i>Panel B: CRAM Data</i>					
Household Level					
<i>Responded to CRAM</i>	0.83	0.01	0.84	0.01	0.89
<i>Household Size</i>	6.20	0.14	6.37	0.15	0.43
<i>Number of Children</i>	2.36	0.09	2.42	0.10	0.66
<i>Urban</i>	0.64	0.02	0.64	0.02	0.94
<i>Receiving Older Person's Grant</i>	0.64	0.02	0.31	0.02	0.00
<i>Receiving Other Government Benefit</i>	0.71	0.02	0.73	0.02	0.42
Respondents (Excluding those around threshold)					
<i>Age</i>	30.95	0.46	31.83	0.48	0.19
<i>Male</i>	0.39	0.03	0.39	0.03	0.93
<i>African</i>	0.90	0.02	0.86	0.02	0.08
<i>Employed Pre-Pandemic</i>	0.40	0.02	0.36	0.02	0.22
<i>Secondary-Level Education</i>	0.51	0.03	0.50	0.03	0.72

Notes: This table shows balance for a Age Range 57-62—Balance is similar for other age ranges considered. This table suggests that households and household members with members just above and just below the Older Person's Grant threshold of age 60 are very similar in the NIDS and the CRAM samples.

dren in the household, if the household is in an urban area, or if they have experienced a death in the last year. Even average non-grant income per capita does not differ between the two groups. We can however, see differences in household-level variables that we expect to change due to the grant, namely, the average grant income per capita, share who have savings, and share who are poor. At the individual level, in our restricted sample, the members above and below the threshold are clearly of different average ages. However, we cannot statistically differentiate between the two groups on the share who are male, married, have secondary-level education, or report a health issue in the last 30 days. When it comes to other members of the household (not including the recipient or potential recipient), their characteristics are similar across the two groups in terms of the age, sex, marital status, and even labor force participation. Similar findings hold for the restricted sample of the CRAM data. Panel B in Table 1 shows strong balance at the household level in response rates, household size, number of children, if the household is in an urban area, and receiving other non-grant government benefits. The only difference that is statistically significant at conventional levels is if the household received the Older Person’s Grant. At the individual level, we again find strong balance, with the exception of whether the individual identifies as African.

After restricting our sample to households with members around the age-eligibility threshold of 60, we discuss two estimation approaches. If receipt of the grant was universal beginning at age 60, we could estimate the simple regression in equation (1) using ordinary least squares:

$$Y_{hdt} = \beta_0 + \beta_1 G_{hdt} + \mathbf{X}'_{hdt} \boldsymbol{\gamma} + \theta_t + \tau_d + \epsilon_{hdt} \quad (1)$$

where Y_{hdt} represents a household-level outcome variable in district d at time t . This variable takes several forms throughout our analysis: (i) household income per capita, (ii) household food expenditures per capita, a wealth index, (iii) whether the household has run out of money for food, and (iv) indicators for adults/child hunger, or (v) psychological distress within the household. The variable G_{hdt} is an indicator of whether an individual within the household receives the Older Person’s Grant and β_1 is our coefficient of interest, giving the relationship between grant receipt and our outcome variables. The vector \mathbf{X}_{hdt} represents household-level control variables that include: household size, number of children, number of elderly, demographics of the household head, and other household level characteristics. Finally, θ_t and τ_d are time, and district fixed effects respectively, and ϵ_{hdt} is the error term.

The coefficient β_1 in equation (1) is potentially biased due to selection into grant receipt. In addition to the age-eligibility requirement, the Older Person’s Grant is means tested such that individuals with earnings or asset holdings above a given threshold are not eligible for the program. Therefore, simply comparing households that receive the grant to those who do not receive the grant, as done in equation (1), would lead to biased estimates of the effects of the grant on indicators of economic and psychological well-being. Therefore, we leverage the age-eligibility threshold of the Older Person’s Grant within an instrumental variable estimation approach. Specifically, we use a dummy variable for having household members who are at least 60 years old as an instrument for grant receipt and estimate the following set of equations:

$$G_{hdt} = \gamma_0 + \gamma_1 I_{hdt} + \mathbf{X}'_{hdt} \Lambda + \delta_t + \phi_d + \zeta_{hdt} \quad (2)$$

$$Y_{hdt} = \alpha_0 + \alpha_1 \hat{G}_{hdt} + \mathbf{X}'_{hdt} \Gamma + \kappa_t + \gamma_d + \mu_{hdt} \quad (3)$$

where I_{hdt} is a variable that indicates the number of household members who are at least 60 years old. The outcome in equation (2), G_{hdt} is an indicator of whether an individual within the household received the Older Person’s Grant. In equation (3) \hat{G}_{hdt} is the predicted value from equation (2). Similar to equation (1), \mathbf{X}_{hdt} is a vector of household level control variables, the equations (2) and (3) each include time and district fixed effects. Finally, ζ_{hdt} and μ_{hdt} are error terms.

We apply this specification on several different age ranges to estimate our pre-pandemic pandemic results using the NIDS data. When the window in local randomization approaches is one, which in our case means restricting our analysis to households members who are only 59 and 60 years. We show this range in our analysis with some caution because it takes several months after turning 60 to apply for and to start receiving the grant and thus (as can be seen in Figure 1) a good portion of 60 year-olds are not yet receiving the grant. Our preferred specifications are those with window sizes of 2 and 3. Table 1 shows balance in the NIDS and CRAM samples with window size 3. Finally, most of our dependent variables of interest are at the household level; however, for one of our results during the COVID-19 pandemic, we estimate the effect of household-level grant receipt on the mental health of the individual who responds to the CRAM phone survey. In 90 percent of the sample, this respondent is not the member who is around the age of 60.

4 Results

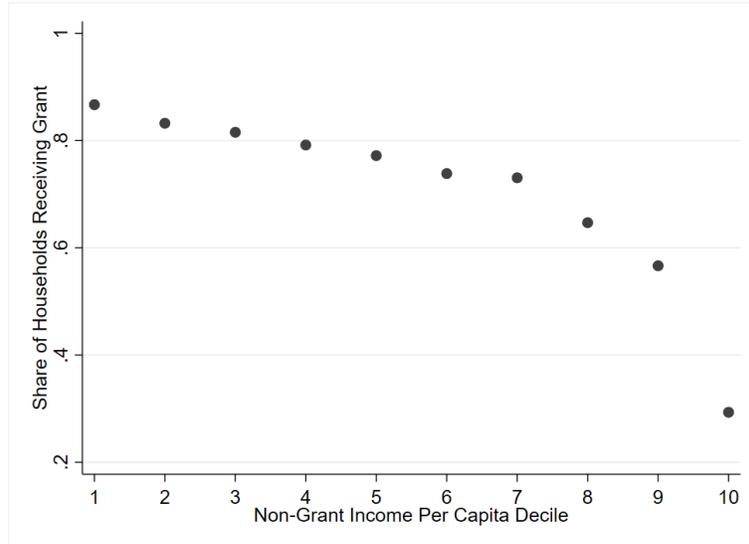
We present three sets of results. First, with over 20 years of new and improved data, we update several stylized facts about the scale and reach of the Older Person’s Grant. These results allow us to demonstrate both which types of households typically receive the grant and the importance of grant income, relative to other forms of income, for recipient households. Second, we document the effect of the grant on indicators of household well-being using pre-pandemic data. Finally, in our core set of results, we report the effect of the grant on key indicators of household well-being during the COVID-19 pandemic.

4.1 Updated Stylized Facts

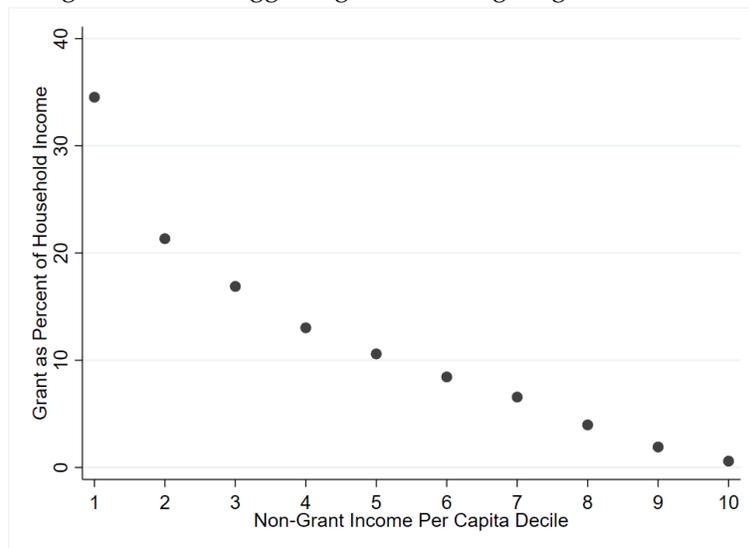
Seminal work by [Case and Deaton \(1998\)](#) describes the scale and scope of South Africa’s Older Person’s Grant program by presenting a number of stylized facts. One of the key descriptive findings reported by [Case and Deaton \(1998\)](#) is that the grant is an effective tool of redistribution as it effectively reaches predominantly poor households. In addition, because many of the elderly in South Africa live with children, the grant is also effective in reaching households where children live, and more specifically where poor children live.

We update these stylized facts using the pre-pandemic NIDS data. In Panel A of [Figure 2](#) we show that over 80 percent of households with a member over the age of 60 in the lowest decile of income per capita receive the Older Person’s Grant and this share declines as non-grant income per capita rises. Panel B of [Figure 2](#) shows that, among all households, the Older Person’s Grant income as a share of total household income declines as non-grant income per capita increases. In particular, the grant represents over 30 percent of total household income for households in the lowest decile of non-grant income per capita. Taken together, these findings demonstrate that South Africa’s Older Person’s Grant continues to reach predominantly poor households and grant income continues to represent an important source of income for poor households. In addition, [Figure A.3](#) in the Supplemental Appendix shows that across all deciles of household wealth, households with children are more likely than households without children to receive the Older Person’s Grant. This is especially true among poorer households, where over one in every three households with children receive the grant. This finding demonstrates that South Africa’s Older Person’s Grant, despite explicitly targeting older recipients, continues to reach households in which children live.

Another finding of [Case and Deaton \(1998\)](#) is that pension income and non-pension income are spent similarly by households. Specifically, the effect of additional pension in-



(A) Among households with a member over 60, the percent of household receiving the Older Person’s Grant decreases with non-grant income suggesting effective targeting.



(B) Among all households, the share of total household income that comes from the grant is decreasing with non-grant income suggesting that among poor households, the Older Person’s Grant makes up a large portion of their well-being.

FIGURE 2: Targeting and Intensity of Treatment

come and non-pension income on food expenditures are both statistically different from zero, but are not statistically different from one another. We update this finding by replicating the analysis of [Case and Deaton \(1998\)](#) with the pre-pandemic waves of the NIDS data. Table [A.1](#) in the Supplemental Appendix replicates the [Case and Deaton \(1998\)](#) results us-

ing their methodological approach.¹⁶ In columns (1) and (2), we estimate the relationship between total income, income excluding grant receipt, and grant income on food expenditure. These regressions control for the number age-eligible adults in the household and show that households spend a larger share of non-grant income relative to grant income on food. This is a qualitatively similar finding to that found by [Case and Deaton \(1998\)](#). In columns (3) and (4), still following [Case and Deaton \(1998\)](#), we instrument for income using the number of age-eligible household members in order to address potential measurement error in the reporting of income.

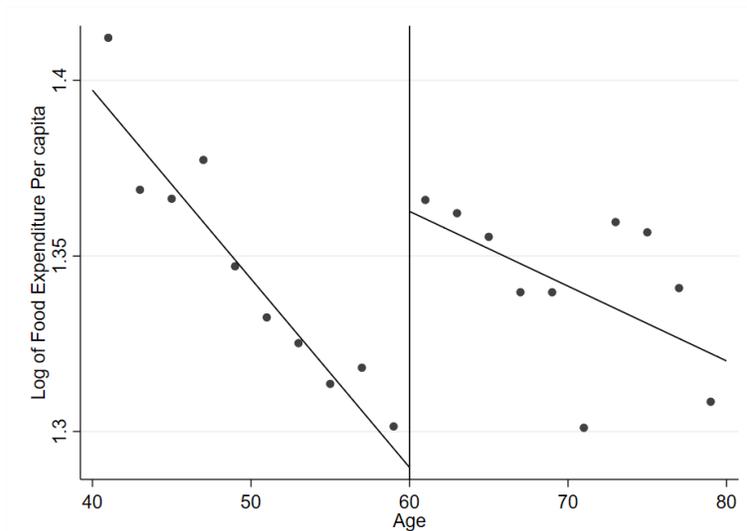
Column (3), which instruments for non-grant income only, returns coefficients for both income sources that are statistically significantly different from zero but not from one another—suggesting that a rand is a rand regardless of its source. Column (4), however, which instruments for both income sources in a manner similar to [Case and Deaton \(1998\)](#)’s preferred specification, returns a larger coefficient on grant income suggesting that more of grant income than non-grant income might go toward food expenditure.

[Case and Deaton \(1998\)](#) describe the limitations of their analysis and suggest how future work could improve these results. In [Table A.2](#) in the Supplemental Appendix, we use our local randomization approach to re-estimate the effect of grant income on food expenditures. We find that, on average, the amount spent on food from the grant is similar in [Table A.2](#) to the results reported in [Table A.1](#). In short, the stylized facts presented by [Case and Deaton \(1998\)](#) 20 years ago continue to reflect the reach and relative scale of the Older Person’s Grant. Not only does a rand of grant income still appear to be as good as a rand of non-grant income when it comes to food expenditure, a rand from the Older Person’s Grant may be even more effective in terms of supporting the socio-economic well-being of poor households.

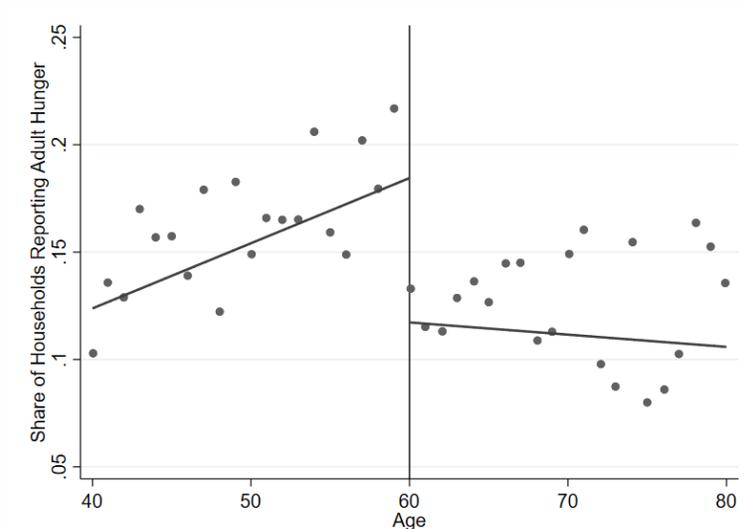
4.2 Pension Receipt Improves Economic Well-Being

We now turn to leveraging the discontinuity in grant receipt to show how important measures of household economic well-being change as a member of the household starts receiving the grant. While there are many documented behavioral changes related to the Older Person’s Grant ([Duflo, 2000, 2003](#); [Bertrand, Mullainathan and Miller, 2003](#); [Ardington, Case and Hosegood, 2009](#); [Lovo, 2011](#); [Edmonds, Mammen and Miller, 2004](#); [Edmonds, 2006](#); [Hamoudi and Thomas, 2014](#); [Abel, 2019](#)), we find that receiving the grant leads to improved economic well-being at the household level.

¹⁶The analogous results are found in [Table 5](#) of ([Case and Deaton, 1998](#))



(A) Household Food Expenditure Per Capita and Age of the Household Head.



(B) Reported Adult Hunger and the Age of the Household Head.

FIGURE 3: Discontinuity in household income and food expenditure per capita by the age of the household head. Pooled sample of all households who met the means test eligibility rule for the Older Person's Grant Program.

Figure 3 illustrates the relationship between age of the household head and key indicators of household well-being. Panel A of Figure 3 uses data from the NIDS and shows that household food expenditures per capita fall gradually as the head of the household ages. Once the household head turns 60 years old, however, we see a sharp increase in household food expenditures. The log of household food expenditures just after the household head turns 60 years old is similar to the log of household food expenditures when the household head is roughly 45 years old.

Although both food expenditures is a useful measure of economic well-being, it is

only instrumentally valuable. We, therefore, further investigate how the age-eligibility threshold influences household well-being by assessing the relationship between reported hunger and the age of the household head. The NIDS does not directly measure hunger, so for this analysis we turn to the 2016 wave of the South African DHS data. In Panel B of Figure 3 we see that hunger increases gradually as the household head ages.¹⁷ As a household head approaches 60 years old, almost 20 percent of households report adult hunger. Once the household head turns 60 years old and is eligible to receive the grant, however, the share of households reporting adult hunger falls to just above 10 percent. In addition, as the household head continues to age, the share of households reporting hunger does not increase. Instead, the Older Person’s Grant seems to keep the rate of hunger relatively consistent or even induce a slight decline—reflecting perhaps that more household members are becoming eligible for the grant. This finding illustrates the important poverty-alleviation role of the Older Person’s Grant. It also supports the validity of our main results in this paper in that using different data we continue to see patterns indicating a strong discontinuity in economic well-being within a narrow age-range around the age-eligibility threshold.

Table 2 presents estimates of the effect of grant receipt on three measures of economic well-being. Using the NIDS data, we apply our local randomization approach with decreasing window size to estimate the effects of grant receipt.¹⁸ In each column in Table 2 we show results with a different window size, from five years to one year on each side of the age-eligibility threshold.¹⁹ Panel A of Table 2 shows that overall household income per capita increases by over 20 percent on average when a member of the household starts to receive the Older Person’s Grant. Panel B of Table 2 shows that food expenditure increases by between six and eight percent at the household-level upon the receipt of the grant by a member of the household. This reinforces work by Case and Deaton (1998) showing that a significant portion of the grant is spent on food. We show in Figure A.4 in the Supplemental Appendix that a larger portion of the grant is spent on food by poorer households,

¹⁷This finding corresponds with the declining food expenditure shown in Panel A and the declining income shown in Figure A.5 in the Supplemental Appendix.

¹⁸Table A.4 in the Supplemental Appendix shows results from a non-fuzzy local randomization approach where our treatment is defined as having a member in the household who is past the eligibility age of 60 but within the bandwidth around the age-eligibility threshold. This estimation approach will understate the effect of the grant because not every individual who is 60 years old or older receives the grant. These are akin to intention-to-treat estimates.

¹⁹We show results with one year on each side of the age-eligibility threshold because doing so is preferred in local randomization literature (Cattaneo, Idrobo and Titiunik, forthcoming). However, given that recipients do not necessarily begin receiving the grant immediately after turning 60 years old, our preferred estimates use a window of two years on each side of the age-eligibility threshold.

TABLE 2: Improvement in Household-Level Economic Well-being Due to Pension

	Member Age Range centered at 60				
	55-64	56-63	57-62	58-61	59-60
<i>Panel A: Log Household Income Per Capita</i>					
<i>Older Person's Grant Receipt</i>	0.235*** (0.036)	0.255*** (0.039)	0.252*** (0.046)	0.241*** (0.056)	0.200** (0.099)
<i>Panel B: Log Food Expenditure Per Capita</i>					
<i>Older Person's Grant Receipt</i>	0.068*** (0.022)	0.067*** (0.024)	0.066** (0.028)	0.078** (0.035)	0.063 (0.067)
<i>Panel C: Wealth Index</i>					
<i>Older Person's Grant Receipt</i>	0.095** (0.039)	0.107** (0.043)	0.061 (0.049)	0.089 (0.061)	-0.037 (0.125)
N	8,312	6,854	5,329	3,654	1,902
Effective First-Stage F-Stat	2,895.9	2,207.6	1,571.9	899.8	339.9

Notes: We instrument for grant receipt using a dummy variable of having a member above the age 60 in the age range reported in the column. We control for wave and district fixed effects. A more restrictive approach focusing on only households with members around age 60 who are economically inactive shows similar results [Alloush and Wu \(forthcoming\)](#). Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

who are the primary intended beneficiaries of the program. Finally, In Panel C we also show that household wealth increases with grant receipt, however, this is not statistically significant for the narrow ranges around age 60 perhaps reflecting that wealth takes time to accumulate.²⁰

4.3 Effects During the COVID-19 Pandemic

Table 3 shows the results of our main local randomization specification applied to the CRAM data. Again, in each column in Table 3 we show results with a different window, from five years to one year on each side of the age-eligibility threshold. Our main specifications follow a local randomization framework where we use having an age eligible member as an instrument for grant receipt. We show first-stage results in Table A.3 in the

²⁰The wealth index is constructed through factor analysis of household-level dwelling characteristics and durable goods (assets).

Supplemental Appendix. These results highlight that our instrument is very strong and predicts grant receipt at the household level. This is despite using household-level information from the prior NIDS survey to determine eligibility of someone in the household for the Older Person's Grant.²¹

In Panel A our instrumental variable estimates show that, for the sample of households that have members in the narrow window around the age of 60, the effect of receiving the Older Person's Grant is a roughly 13 percentage point reduction in the likelihood of running out of money to buy food.²² Panels B and C show that household-level adult and child hunger in the seven days prior to the interview are, in our preferred specifications with two years on each side of the age-eligibility threshold, between 11 and 14 percentage points lower among households who receive the grant. A weighted average of all five coefficients gives an estimate of 10 percentage point reduction in reported adult hunger and a 7 reduction in reported child hunger. With rates of adult hunger at 26 percent and of child hunger at 15 percent (Wills et al., 2020), our estimates imply that the Older Person's Grant led to a nearly 40 percent reduction in adult hunger and a nearly 45 percent reduction in child hunger during the COVID-19 pandemic.²³ Figure A.6 in the Supplemental Appendix shows reported adult hunger in our narrow sample (member between age 55 and 64) for households with members just above and just below the threshold. In addition, we show the average reported hunger for other households. Those with a member above the threshold consistently have lower levels of reported hunger throughout all five CRAM waves.²⁴ Panel D reports supplemental results showing that grant receipt reduces "extreme hunger" which is defined as respondents reporting that someone in their household has to eat less than they would like almost daily.

Finally, in Panel E, we use a measure of psychological distress as recorded by the CRAM survey. This measure reports whether the individual CRAM survey respondent

²¹The CRAM was designed as a rapid phone survey and sampled from adults who part of the fifth wave of NIDS—however, detailed household information was not collected in the CRAM survey. We used household-level information from the fifth wave of the NIDS and projected the household members' ages forward to predict who would be eligible for the Older Person's Grant.

²²This number is the weighted average of the all five coefficients. The average is 13.5 when averaging the coefficients estimated with the middle three age ranges.

²³Weighted averages of the coefficients estimated with the middle three age ranges give estimates of 9.3 and 6.2 reduction in hunger among adults and children, respectively implying slightly lower percent reductions in hunger rates.

²⁴Given that our pre-pandemic measure of hunger comes from the DHS data and uses a slightly different method of measuring hunger than used in the CRAM data collected during the pandemic, we cannot make an easy comparison of this outcome across these two sets of data. With that said, when we apply the local randomization estimation approach to the DHS data, we find that grant receipt leads to about a 7 percent decline in hunger.

TABLE 3: COVID-19 and Older Person’s Grant Receipt

	Member Age Range centered at 60				
	55-64	56-63	57-62	58-61	59-60
<i>Panel A: Report Running out of Money for Food</i>					
<i>Older Person’s Grant Receipt</i>	-0.081** (0.036)	-0.125*** (0.040)	-0.111** (0.048)	-0.199*** (0.063)	-0.259 (0.118)
<i>Panel B: Report Adult Hunger</i>					
<i>Older Person’s Grant Receipt</i>	-0.082*** (0.029)	-0.078** (0.032)	-0.083** (0.039)	-0.135*** (0.051)	-0.276*** (0.096)
<i>Panel C: Report Child Hunger</i>					
<i>Older Person’s Grant Receipt</i>	-0.083*** (0.025)	-0.057** (0.027)	-0.031 (0.033)	-0.115*** (0.044)	-0.099 (0.078)
<i>Panel D: Extreme Hunger (Almost Daily)</i>					
<i>Older Person’s Grant Receipt</i>	-0.059** (0.023)	-0.075*** (0.026)	-0.081*** (0.032)	-0.094** (0.042)	-0.076 (0.077)
<i>Panel E: Psychological Distress</i>					
<i>Older Person’s Grant Receipt</i>	-0.050 (0.046)	-0.059 (0.050)	-0.067 (0.062)	-0.161** (0.081)	-0.369** (0.158)
N	6,045	5,095	3,923	2,704	1,240
Effective First-Stage F-Stat	954.9	778.5	518.4	290.8	78.9

Notes: We control for wave and district fixed effects in addition to a host of household controls. We instrument for grant receipt using a dummy variable of having a member above the age 60 in the age range reported in the column. Panel E is the reported psychological distress of the member who is responding to the CRAM phone-interview who in over 90% of the observations is under the age of 55 and is not the recipient or potential recipient of the Older Person’s Grant. Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

had experienced psychological distress in the past month. We find that having a household member who is eligible for and is receiving the Older Person’s Grant led to a reduction in reported psychological distress during the COVID-19 pandemic. Although estimates of these effects using our local randomization approach are not all statistically significant, the average effects remain relatively large. Specifically, a weighted average of the coefficients estimated with the middle three age ranges suggests that grant receipt leads to an 9

percentage point decline in the likelihood the survey respondent experienced psychological distress in the past month—a nearly 25 percent reduction. Given that the CRAM survey respondent is usually not the actual pensioner, these results suggest that the grant has psychological benefits on other members of the household beyond the previously discussed economic benefits. Moreover, this result is important given the high levels of psychological distress documented during the pandemic.

Table A.5 in the Supplemental Appendix shows results from analysis where our treatment is simply defined as having a member in the household who is past the eligibility age of 60 but within our relatively narrow bandwidth around the age-eligibility threshold. This estimation approach will, by definition, understate the effect of the grant because not every individual who is 60 years old or older receives the grant. These results supplement the core results presented in the main manuscript and are akin to intention-to-treat estimates.

5 Conclusion

The COVID-19 pandemic hit South Africa early and hard. With nearly half of the population vulnerable and living in poverty, the economic disruptions caused by the pandemic resulted in high levels of hunger and psychological distress. Our paper shows that a well-targeted unconditional cash transfer program—the Older Person’s Grant—played an important role in allowing recipient household to better manage the adverse consequences of a global health crisis and the associated lockdowns.

The Older Person’s Grant has a wide reach in South Africa and constitutes a large portion of the overall net income of poor households. Prior to the pandemic, the program significantly improved the economic well-being of recipient households and reduced reported hunger. During the COVID-19 pandemic, this reliable source of income is linked with between 40 and 45 percent lower rates of adult and child hunger in the household. In addition, individuals living in households with a grant recipient were less likely to report psychological distress.

These results provide important insight into the effectiveness of large cash transfer programs around the world at helping households manage large and unexpected global shocks. Many developing countries have instituted or expanded these types of programs in response to the COVID-19 pandemic. Further, interest in large cash transfer programs is not limited to developing countries nor to acute disaster response. A key feature of South Africa’s Older Person’s Grant is that it has been providing a reliable source of income for

decades, allowing individuals to confidently incorporate this source of income into their response to shocks. Wealthier countries are increasingly looking to build similarly targeted and reliable instruments into their social safety programs—for example in the form of tax credits for low income households with children in the United States. The South African example suggests that these programs can have important effects on the resilience and well-being of both the target population and those close to them.

References

- Abay, Kibrom, Guush Berhane, John Hoddinott, and Kibrom Tafere.** 2021. "COVID-19 and food security in Ethiopia: Do social protection programs protect?" *Economic Development and Cultural Change*, forthcoming.
- Abel, Martin.** 2019. "Unintended Labour Supply Effects of Cash Transfer Programmes : Evidence from South Africa ' s Old Age Pension." *Journal of African Economies*, 28(5): 558–581.
- Alloush, M., and Stephen Wu.** forthcoming. "Income Improves Subjective Well-being: Evidence from South Africa." *Economic Development and Cultural Change*.
- Ardington, C., A. Case, and V. Hosegood.** 2009. "Labor Supply Responses to Large Social Transfers: Longitudinal Evidence from South Africa." *American Economic Journal: Applied Economics*, 1(1): 22–48.
- Arellano, Manuel, and Stephen Bond.** 1991. "Some tests of specification for panel data: Monte Carlo evidence and an application to employment equations." *The Review of Economic Studies*, 58(2): 277.
- Arndt, Channing, Rob Davies, Sherwin Gabriel, Laurence Harris, Konstantin Makrelov, Sherman Robinson, Stephanie Levy, Witness Simbanegavi, Dirk van Seventer, and Lillian Anderson.** 2020. "Covid-19 lockdowns, income distribution, and food security: An analysis for South Africa." *Global Food Security*, 26: 100410.
- Banks, J., R. Blundell, and A. Lewbel.** 1997. "Quadratic Engle Curves and Consumer Demand." *Review of Economics and Statistics*, 79(4): 527–539.
- Bertrand, Marianne, Sendhil Mullainathan, and Douglas Miller.** 2003. "Public policy and extended families: Evidence from pensions in South Africa." *the world bank economic review*, 17(1): 27–50.
- Case, Anne, and Angus Deaton.** 1998. "Large Cash Transfers to the Elderly in South Africa." *The Economic Journal*, 108: 1330–1361.
- Cattaneo, Matias D, Nicolás Idrobo, and Rocío Titiunik.** forthcoming. *A practical introduction to regression discontinuity designs: Extensions*. Cambridge University Press.
- Duflo, Esther.** 2000. "Child health and household resources in South Africa: evidence from the old age pension program." *American Economic Review*, 90(2): 393–398.

- Duflo, Esther.** 2003. "Grandmothers and granddaughters: old-age pensions and intra-household allocation in South Africa." *The World Bank Economic Review*, 17(1): 1–25.
- Edmonds, E., K. Mammen, and D.L Miller.** 2004. "Rearranging the Family? Income Support and the Elderly Living Arrangements in a Low-Income Country." *Journal of Human Resources*, 40(1): 186–207.
- Edmonds, Eric V.** 2006. "Child labor and schooling responses to anticipated income in South Africa." *Journal of Development Economics*, 81(2): 386–414.
- Gentilini, Ugo, Mohamed Almenfi, John Blomquist, Pamela Dale, Luciana De la Flor Giuffra, Vyjayanti Desai, Maria Belen Fontenez, Guillermo Galicia, Veronica Lopez, Georgina Marin, Ingrid Veronica Mujica, Harish Natarajan, David Newhouse, Robert Palacios, Ana Patricia Quiroz, Claudia Rodriguez Alas, Gayatri Sabharwal, and Michael Weber.** 2021. "Social protection and jobs responses to COVID-19."
- Goodman, Peter S., Abdi Latif Dahir, and Karan Deep Singh.** 2020. "The Other Way COVID Will Kill: Hunger." *New York Times*.
- Gulesci, Selim, Manuela Puente-Beccar, and Diego Ubfal.** 2021. "Can Youth Empowerment Programs Reduce Violence against Girls during the COVID-19 Pandemic?" *Journal of Development Economics*, 153.
- Hamoudi, Amar, and Duncan Thomas.** 2014. "Endogenous coresidence and program incidence: South Africa's old age pension." *Journal of development economics*, 109: 30–37.
- Londoño-Vélez, Juliana, and Pablo Querubin.** 2022. "The impact of emergency cash assistance in a pandemic: Experimental evidence from Colombia." *Review of Economics and Statistics*, 104(1): 157–165.
- Lovo, Stefania.** 2011. "Pension Transfers and farm household technical efficiency: Evidence from South Africa." *American Journal of Agricultural Economics*, 93(5): 1391–1405.
- Miguel, Edward, and Ahmed Mushfiq Mobarak.** 2021. "The Economics of the COVID-19 Pandemic in Poor Countries." *NBER Working Paper*, No. 29339.
- Olea, José Luis Montiel, and Carolin Pflueger.** 2013. "A robust test for weak instruments." *Journal of Business & Economic Statistics*, 31(3): 358–369.
- Van der Berg, Servaas.** 1997. "South African social security under apartheid and beyond." *Development Southern Africa*, 14(4): 481–503.

Wills, Gabrielle, Servaas Van der Berg, Leila Patel, Bokang Mpeta, et al. 2020. *Household resource flows and food poverty during South Africa's lockdown: Short-term policy implications for three channels of social protection*. Department of Economics, University of Stellenbosch.

Appendix

Tables

TABLE A.1: Comparing Sources of Income and Food Expenditure

	Case & Deaton (1998) Approach			
	Pooled		IV	
	OLS			
	(1)	(2)	(3)	(4)
Total Income	0.035** (0.014)			
Income Excluding Grant		0.035** (0.014)	0.070*** (0.004)	0.073*** (0.004)
Grant Income		-0.060 (0.045)	0.090*** (0.016)	0.169*** (0.020)
Number age-eligible females		1.615*** (0.432)		
Number age-eligible males		1.411*** (0.428)		
<i>Controls</i>	✓	✓	✓	✓
N	41,196	41,196	41,196	41,196

Notes: In Column (3), we instrument for non-grant income using employment and individual sex and race characteristics replicating the specification of [Case and Deaton \(1998\)](#). In Column (4), we additionally instrument for Grant income using the number of eligible elderly (above age 60) in the household also replicating the specifications used by [Case and Deaton \(1998\)](#). In Columns (5) and (6), we take advantage of the panel nature of the data and re-estimate the main regressions in [Case and Deaton \(1998\)](#) using a GMM approach that attempts to take into account household fixed effects and other endogeneity issues [Arellano and Bond \(1991\)](#); [Banks, Blundell and Lewbel \(1997\)](#). Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

TABLE A.2: Food Expenditure and the Older Person's Grant

		OLS		IV	
		(1)	(2)	(3)	(4)
Age Range					
54-65	<i>Grant Income</i>	0.096***	0.060***	0.168***	0.118***
N=9,438		(0.014)	(0.013)	(0.025)	(0.028)
Age Range					
55-64	<i>Grant Income</i>	0.097***	0.067***	0.167***	0.123***
N=8,062		(0.014)	(0.013)	(0.027)	(0.030)
Age Range					
56-63	<i>Grant Income</i>	0.094***	0.062***	0.165***	0.122***
N=6,651		(0.016)	(0.014)	(0.033)	(0.034)
Age Range					
57-62	<i>Grant Income</i>	0.089***	0.058***	0.147***	0.098**
N=5,171		(0.020)	(0.016)	(0.040)	(0.042)
Age Range					
58-61	<i>Grant Income</i>	0.105***	0.075***	0.162***	0.120**
N=3,542		(0.025)	(0.017)	(0.060)	(0.061)
Age Range					
59-60	<i>Grant Income</i>	0.107***	0.069***	0.140	0.122
N=1,841		(0.038)	(0.026)	(0.107)	(0.104)
Controls			✓		✓

Notes: The results in this table re-estimate the main results in [Case and Deaton \(1998\)](#) with a local randomization approach and specific samples to take into account several sources of endogeneity pointed out in their work. Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** p<0.01, ** p<0.05, * p<0.1.

TABLE A.3: First Stage regression results

Dependent Variable: A member of the household receives the Older Person's Grant		IV First Stage	Effective F-Stat
Age Range 55-64 N=5,676	<i>Number of HH members 60-64</i>	0.357*** (0.012)	860.3
Age Range 56-63 N=4,785	<i>Number of HH members 60-63</i>	0.353*** (0.013)	778.5
Age Range 57-62 N=3,688	<i>Number of HH members 60-62</i>	0.338*** (0.015)	518.4
Age Range 58-61 N=2,342	<i>Number of HH members 60-61</i>	0.314*** (0.018)	290.8
Age Range 59-60 N=1,171	<i>Number of HH members 60</i>	0.290*** (0.033)	78.9

Notes: First stages of the IV regressions show very strong predictive value of the instrumental variable for the log of household income. These first stage results correspond to the IV regression results in Table 3 column (5). Effective F-statistics according to [Olea and Pflueger \(2013\)](#) are shown. Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

TABLE A.4: Household Economic Well-being and Grant Age-Eligibility

	Member Age Range centered at 60				
	55-64	56-63	57-62	58-61	59-60
<i>Panel A: Log Household Income Per Capita</i>					
<i>Member above 60</i>	0.133*** (0.020)	0.135*** (0.020)	0.124*** (0.022)	0.105*** (0.024)	0.073** (0.036)
<i>Panel B: Log Food Expenditure Per Capita</i>					
<i>Member above 60</i>	0.039*** (0.012)	0.035*** (0.013)	0.032** (0.014)	0.034** (0.015)	0.023 (0.024)
<i>Panel C: Wealth Index</i>					
<i>Member above 60</i>	0.055** (0.023)	0.059** (0.023)	0.031 (0.025)	0.040 (0.028)	-0.014 (0.046)
N	8,312	6,854	5,329	3,654	1,902

Notes: These results are estimated using ordinary least squares regressions. We control for wave and district fixed effects. These are akin to ITT estimates for Table 2 where instead of using having a member as an instrument for grant receipt, we simply regress our economic well-being measures on having a member above 60 but restricted to our age windows. Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

TABLE A.5: COVID-19 and Grant Age-Eligibility

	Member Age Range centered at 60				
	55-64	56-63	57-62	58-61	59-60
<i>Panel A: Report Running out of Money for Food</i>					
<i>Member above age 60</i>	-0.029** (0.013)	-0.044*** (0.014)	-0.038** (0.016)	-0.062*** (0.020)	-0.075** (0.034)
<i>Panel B: Report Adult Hunger</i>					
<i>Member above age 60</i>	-0.029*** (0.010)	-0.027** (0.011)	-0.028** (0.013)	-0.042*** (0.016)	-0.080*** (0.027)
<i>Panel C: Report Child Hunger</i>					
<i>Member above age 60</i>	-0.030*** (0.009)	-0.020** (0.010)	-0.010 (0.011)	-0.035*** (0.014)	-0.029 (0.022)
<i>Panel D: Extreme Hunger (Almost Daily)</i>					
<i>Member above age 60</i>	-0.021** (0.008)	-0.026*** (0.009)	-0.027** (0.011)	-0.019** (0.013)	-0.022 (0.022)
<i>Panel E: Psychological Distress</i>					
<i>Member above age 60</i>	-0.017 (0.016)	-0.021 (0.018)	-0.022 (0.020)	-0.050** (0.025)	-0.101** (0.043)
N	6,045	5,095	3,923	2,704	1,240
Effective First-Stage F-Stat	960.2	744.1	746.0	503.9	216.0

Notes: We control for wave and district fixed effects in addition to a host of household controls. These results show differences within the window for households with members above age 60 compared to those with members below age 60; this is akin to an intent-to-treat regression as being eligible does not mean that the member is receiving the Older Person's Grant. Standard errors clustered at the original (i.e., wave 1) sampling cluster area are presented in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Figures

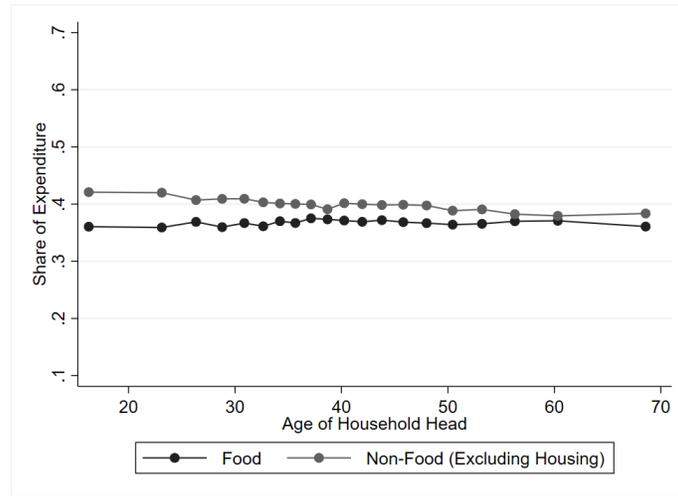


FIGURE A.1: Share of income spent on food by age of the household head. This figure suggests that preferences regarding food expenditure do not change abruptly at age 60.

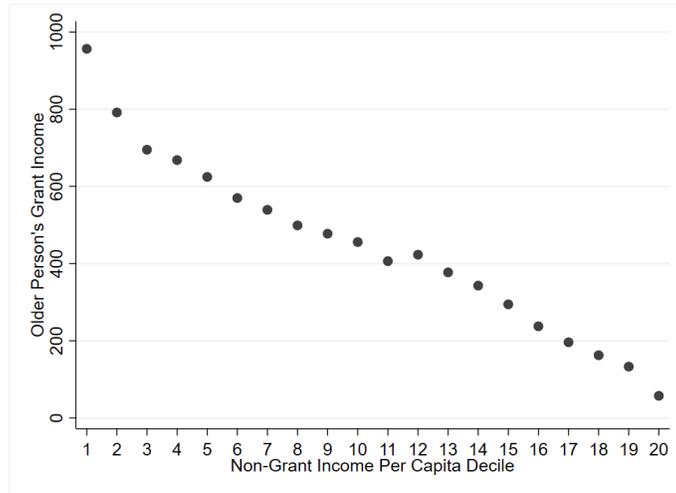


FIGURE A.2: Average amount of grant received by households by non-grant income per capita. This figure shows results for the full Sample.

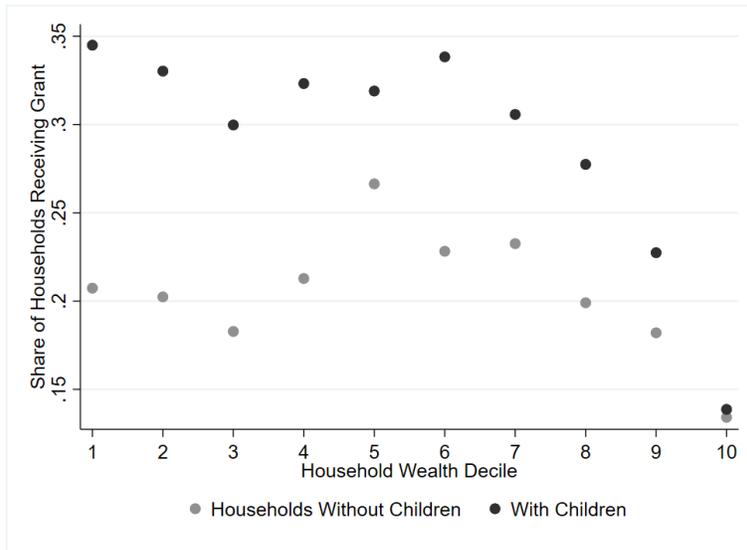


FIGURE A.3: Wealth deciles and grant receipt for households with and without children.

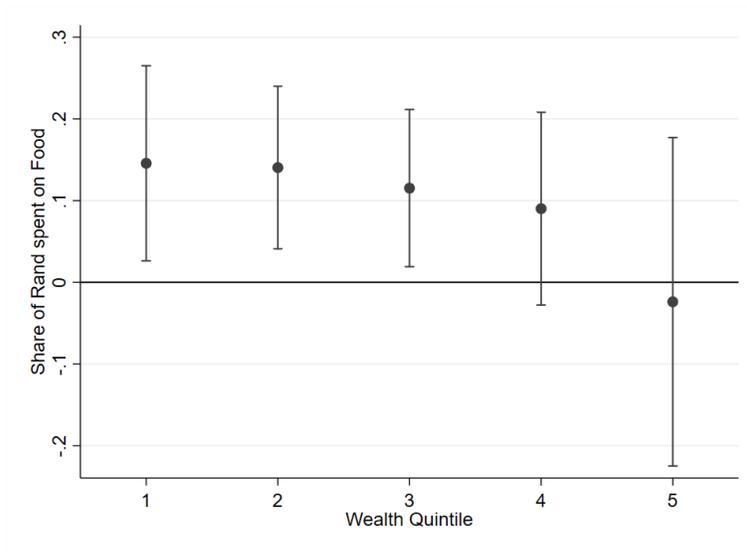


FIGURE A.4: Portion of Grant spent on food by wealth.

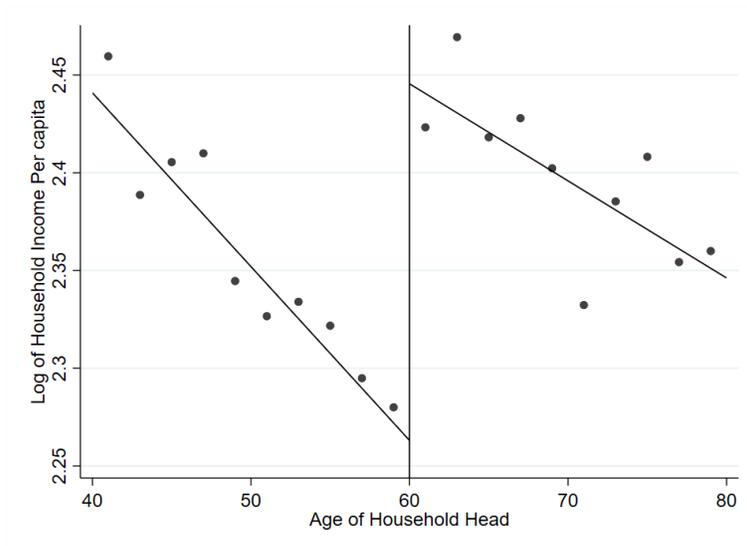


FIGURE A.5: Discontinuity in household income per capita by the age of the household head. Pooled sample of all households who met the means test eligibility rule for the Older Person's Grant Program.

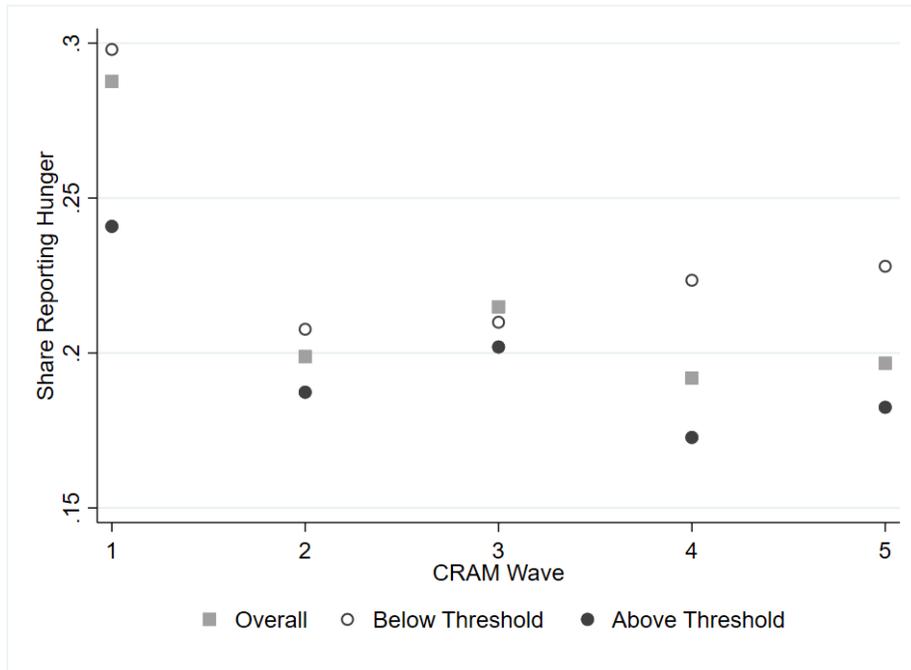


FIGURE A.6: Reported Adult Hunger in a small window around Old Age Grant Receipt during Covid-19 Pandemic