Information Dissemination and Active Citizenship: 
Experimental Evidence from Uganda

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Abstract

Information constraints are seen as serious impediments to the ability of citizens to hold politicians accountable. In this thesis, I study an intervention in Uganda where scorecards describing the performance of local politicians are disseminated to the constituencies of randomly selected local politicians. I test the effects of the scorecard dissemination on citizens’ participation and demand for improved public services. I find that the information has no effect on the number of citizens reporting service delivery problems and that the effect is negative on the number citizens attending community meetings and speaking at community meetings called by local governments. By studying a unique set of participation outcomes, the thesis contributes to the scarce literature investigating the link between information about politicians and citizen participation in forms other than voting.

Keywords: Social Accountability, Participation, Information, Public Service Delivery, Difference-in-Difference, Propensity Score Matching
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1. Introduction

The poor and vulnerable often do not get much attention from politicians, partly because they are less informed and generally less inclined to participate politically than citizens higher up on the socioeconomic ladder (Besley and Burgess, 2002). This is especially problematic since this group is the most dependent on public services and hence has the most to gain from holding politicians and service providers accountable (Malena et al., 2004).

One of the main identified obstacles to the ability of citizens to hold politicians accountable is information asymmetries (cf. Besley and Burgess, 2002; Devarajan et al., 2013; Paul, 1992). When citizens lack information, they can with difficulty evaluate the quality and efficiency of public service delivery and their ability to hold governments accountable is reduced (Keefer and Khemani, 2005). Correcting information asymmetries through information campaigns has been found to increase voter turnout (Banerjee et al., 2011; Strömberg, 2004), improve public service delivery (Bjorkman and Svensson, 2009; Casey et al., 2011), reduce leakages of public funds (Reinikka and Svensson, 2005) and increase government responsiveness (Besley and Burgess, 2002). However, as numerous scholars have shown, providing citizens with information does not always result in (positive) actions being taken by citizens (Banerjee et al., 2010; Chong et al., 2010; Lieberman et al., 2014), causing uncertainty about the widespread assumption that information will leverage accountability through increased pressure from citizens (Fox, 2015).

In this thesis, I test the effects of providing citizens with information about the performance of their locally elected politicians (district councilors) and ask whether such information induces more citizens to actively participate and demand better public services. To address this question, I exploit a unique randomized field experiment in Uganda creating geographical variations in citizens’ exposure to scorecards describing how well district councilors perform their legally defined duties. The scorecard dissemination was designed to make the scorecards common knowledge within the constituencies of randomly selected district councilors, thereby creating an opportunity to make comparisons between individuals in constituencies where the scorecards were disseminated and individuals in constituencies where they were not.
Drawing on individual level data from the Secure Livelihoods Research Consortium (SLRC) Panel Survey, I create six participation outcomes. The first three measurements concern whether an individual has attended any community meeting in which a public service was discussed, spoken at such a community meeting and reported a service delivery problem to any actor. The last three measurements are alternative versions of the first three and only concern actions more directly connected to local governments: attending a community meeting and speaking at a community meeting called by local government and reporting a service delivery problem to local government. I refer to the first three measurements as general participation outcomes and the last three as local government participation outcomes.

The foundation for the experiment (scorecard dissemination) was laid out in 2009 when a Ugandan civil society organization (CSO), ACODE, initiated a program named the Local Government Councils Scorecard Initiative (LGCSCI) and began constructing annual scorecards describing the performance of local politicians in a few selected districts. The experiment was conducted in 20 districts where ACODE implements LGCSCI, meaning that individuals in these districts were randomly assigned to either a treatment group or a control group. Since the 20 districts partly were selected to achieve national representativeness, only 4 districts found in the SLRC dataset (covering northern Uganda) were subject to the experiment. Hence, exploiting the full sample means that treatment is unlikely to be random. Assessing the effects of the scorecard dissemination therefore necessitates resorting to empirical methods accounting for this potential non-randomness. The main method used in this thesis is difference-in-difference (DID), which has the advantage that it accounts for unobservable fixed effects. I employ Propensity Score Matching (PSM) as a robustness check, which I argue is less credible in this context since data limitations could make the fulfillment of the main identifying assumption (conditional independence assumption) unlikely. In addition to assessing the impact as a whole, the fact that whole geographical areas are considered treated allows me to test for heterogeneous treatment effects on individuals with different characteristics and attitudes.

The main finding, using the DID method, is that the scorecard dissemination has negative effects on the number of people attending community meetings and speaking at community meetings, but only for meetings called by local governments. The results using PSM, however, partly contradicts these findings. While both methods fail to find any treatment effects for the more general forms of participation, the PSM results show that, among the local government participation outcomes, reporting a service delivery problem is the only one
on which the treatment effect is negative and significant. In addition, I find evidence of some heterogeneity in treatment impact, but only for the local government participation outcomes. Individuals who believe that local governments are responsive to their opinions are more likely to attend community meetings and speak at community meetings. In contrast, individuals that are dissatisfied with government priorities are more likely to participate in all three forms.

The contributions of the thesis are threefold. First, there is an ongoing debate on whether information by itself is sufficient or if it needs to be accompanied by other interventions, such as the creation of channels through which citizens can use their voice, to instigate behavioral changes and strengthen accountability. Few papers have investigated campaigns that solely rely on information dissemination. Banerjee et al (2010) find that providing information alone is not sufficient while Barr et al. (2012) shows positive effects of disseminating scorecards but that the effects are larger if the construction of the scorecard is a participatory process where the community is involved. Both these papers study the education sector. Bruns et al. (2011) point out that information in certain settings needs to be accompanied by other interventions in order to escape the status quo of citizen disengagement in public services, while it in other settings might be sufficient to solely provide citizens with information to increase their participation rates. They go on to conclude that, if effective, solely relying on information is a cost-efficient approach to strengthen accountability and that research into that area is likely to be valuable. The intervention studied in this thesis was unaccompanied by civic education and did not provide any information on how citizens could take action. The findings of the thesis hence lend support to the literature arguing that information by itself is not enough.

Second, there are very few papers that have investigated the effects of informing citizens about politicians’ performance on outcomes other than related to voting (e.g. voter turnout, voting for the incumbent). Within the literature focusing on developing countries, I only managed to find two such papers. Krishna (2006) investigates the relationship between citizens’ access to information and an index of participation which includes campaigning, attending public meetings and voting. Gottlieb (2016) looks at the effect of information, in the form of a civic education course, on citizens’ challenges to politicians in town hall meeting. Both these papers differ from my thesis in the form of information considered and the participation outcomes. Non-institutionalized forms of participation, such as participating in meetings, are by researchers increasingly seen as an important link through which formal political outcomes can be influenced (Coffé and Bolzendahl, 2010). Contributions to the
scarce literature investigating whether information can have an effect on such forms of participation could hence be of value.

Third, the literature examining the effects of information related to public services on citizen engagement has largely focused on one single public service, such as education (Banerjee et al., 2010; Bruns et al., 2011; Lieberman et al., 2014; Reinikka and Svensson, 2005), health (Bjorkman and Svensson, 2009) and roads (Olken, 2007), overlooking the important role of governments to monitor service delivery and ensure the sufficient quality of services. To the best of my knowledge, no previous research has investigated the effect of information that can be tied to politicians’ performance in ensuring the quality of public services on citizens’ efforts to hold politicians accountable for their performance in this respect. The field experiment studied in this thesis, as well as the Ugandan context, provide a good opportunity to investigate this effect for three reasons: local governments in Uganda are responsible for the delivery of most major services (Mitchinson, 2003), it is the legally defined duty of local politicians to monitor these services and monitoring service delivery makes up 45% of the total scorecard score.

The thesis proceeds as follows. Section 2 gives an overview of the Ugandan context and describes the intervention studied. Section 3 provides the theoretical framework of the thesis, reviews the previous literature and formulates the hypotheses. Section 4 describes the data used and the constructed measurements of participation. Section 5 describes the research design and the empirical strategy. Section 6 reports the results and section 7 concludes.

2. Context and Intervention

2.1 The Ugandan Context

When the National Resistance Movement (NRM) led by Yoweri Museveni came to power in 1986 after 5 years of civil war, it was by many seen as a welcome hiatus from chaos (Tripp, 2010). At that time, the economy was in ruins, public service delivery was insufficient and corruption was widespread (Meyers, 2014). Under the NRM, Uganda has seen a period of relative stability with increased market liberalization through the removal of price controls, high economic growth and reduced poverty, which has contributed to Uganda being seen as a donor-darling (Francis and James, 2003).
Brief history of local government in Uganda

In 1992, NRM initiated a far-reaching decentralization reform which was adopted into legislation in the 1995 constitution and the 1997 Local Government Act (LGA). With these acts, local councils were given extensive responsibilities and powers in areas such as revenue collection, legislation, planning and budgeting and service delivery (Green, 2010). At that time, NRM saw local councils as an important component in its strategy to build democracy from the grassroots and increase citizen participation (Tripp, 2010). Recently, however, scholars have argued that the decentralization process has been captured by NRM to reaffirm control. Key positions at district level are appointed by the central government, often based on loyalty rather than merit, serving as an extended arm of government (Lewis, 2014). More opportunities to appoint NRM loyalist to key positions within local governments have also arisen with the rapid proliferation of districts (112 today compared to 33 in 1990) enabling the government to expand its patronage networks (Booth et al., 2014; Green, 2010).

Local government structure

The local government structure in Uganda consists of 5 tiers: district (LC5), county (LC4), sub-county (LC3), parish (LC2) and village (LC1). Districts and sub-counties are corporate bodies, similarly structured with a local parliament of elected representatives. The remaining levels are administrative units with no focal role in delivering public services (Steiner, 2006). The district council (DC), which is the governing body of interest to this thesis, consists of councilors representing each sub-county within a district and is headed by a chairman, equivalent to a governor in many countries (Saito, 2003). The councilors and the chairman are elected at the local government elections by universal adult suffrage. The DC also consists of administrative staff appointed by the central government.

In addition to directly elected (regular) councilors, a position of “woman councilor” has been established in order to achieve the statutory requirement that a third of the district councilors should be female. Woman councilors represent one to three sub-counties, depending on population size, and are the units to which the scorecard dissemination was randomized (Grossman and Michelitch, 2018).

Local governments are responsible for a wide range of public services in Uganda, such as health services except for referral hospitals; feeder roads; water, education, except tertiary; land administration; and extension services (Awortwi, 2011). In line with the principle of
subsidiary, the idea is that the lowest tier of government capable of delivering a service should be charged with doing so (Steiner, 2008).

Media and Civil Society

The media and civil society has enjoyed relative freedom in Uganda under the NRM regime compared to previous regimes and other countries in the region (Booth et al., 2014). More recently, however, the number of newspapers has declined drastically and the NRM is today far less accepting of direct criticism than upon coming to power, using threats of revoking media licenses and taking legal measures towards individual journalists as means to silence the critical voices (Booth et al., 2014). Similarly, NGOs focusing on issues interfering with the government’s agenda face the threat of de-registration and closure. NGOs focusing on development issues, however, have been welcomed by the government, reflecting its reliance on NGOs to improve social development and service delivery, which in turn indirectly strengthens the legitimacy of the government. NGOs and media addressing issues at the lower tiers of government also tend to be less controversial and are allowed to operate more freely (Sjögren, 2013).

The main media consumed in Uganda is local radio which plays an important role in strengthening local accountability. Local radio stations inform citizens about local issues and regularly hold phone-in programs where citizens can call to ask questions and issue complaints to invited local politicians (Devas and Grant, 2003). In the 2014 census 55% of respondents named radio as their primary source of information, which largely surpassed other sources such as word of mouth (20%), Internet (7.3%), TV (7.2%) and print media (2.1%) (UBOS, 2016). Unlike newspapers and TV, which are almost exclusively in English and are unaffordable by many, radio stations have a large potential to reach the rural poor (Chibita and Fourie, 2007)

Civic Participation

While the local government system was designed to promote citizens’ participation in local development and strengthen the accountability relationship between citizens and politicians, these objectives have only been achieved to a limited extent (Kiyaga-Nsubuga and Olum, 2009).

The low information environment has been a serious impediment to citizens’ participation in Uganda. While citizens are largely dissatisfied with the state of service delivery and corruption in the public sector, they are often unaware of how to confront these issues
Deininger and Mpuga, 2005). Citizens also tend to be unaware of who is responsible and the source of financing for local projects that benefit the communities. This creates space for local politicians to take credit for these projects and generate a perception that it is them and not agencies, NGOs or local governments, that bring improved public services through their support and lobbying (Steiner, 2008).

Low levels of development in Uganda serve as additional constraints to citizen participation. Poor citizens might be reluctant to participate in community meetings due to the high opportunity costs in terms of forgone income, the incomprehensibility of issues discussed and the little resources at stake (Steiner, 2007). The possibility for citizens to affect allocations of funds at the local level is limited. First, a high proportion, if not all, of locally generated revenues go to councilor salaries. Second, the government grants, which constitute the main share of the local budgets, are often subject to conditions which limits the discretion of local governments (Francis and James, 2003). It may hence seem futile for citizens to attempt to influence local budget allocations.

Cultural factors may equally prevent citizens from participating. Among Ugandan citizens there is an expectation that it is the elite groups and not the poor that should participate in local development (Steiner, 2008). Public services are also often seen as personalized and believed to be used as patronage to citizens in the same political camp or network of politicians (Titeca, 2006). Citizens identifying with other groups may hence be discouraged from voicing their concerns over public services. The role of the state in delivering services might also not be fully understood by citizens due to Uganda’s history of poor or nonexistent public services. Saito (2003) notes that health problems are widely believed to be a family problem rather than a community issue in Uganda, which may prevent people from collectively seeking to improve health services.

2.2 Intervention and treatment

Background
The Local Government Councils Scorecard Initiative (LGCSCI) is implemented by a Ugandan CSO, ACODE, in partnership with Uganda Local Governments Association (ULGA). The initiative is financed by a broad based coalition of donors through the
Democratic Governance Facility (DGF) program. Within the framework of LGCSCI, ACODE has since 2009 constructed annual scorecards describing how well local leaders have performed the responsibilities vested in them under the Local Government Act (LGA). The scorecard implementation districts were selected by ACODE to ensure a mix of geographical areas, old and new districts, rich and poor districts and influential and marginalized districts (Bainomugisha et al., 2017).

The scorecard

According to the LGA of 1997, the elected representatives of the district council (DC) have four legally defined duties: participating with lower tier governments (e.g. attending meetings, forwarding issues to DC), contact with the electorate (e.g. meeting with electorate, setting up public office), monitoring public service delivery (e.g. visits to service delivery units, preparing reports) and legislative duties (e.g. participating in committees, moving motions)\(^1\). Each legally defined area is assessed by various parameters given weights so that the total score adds up to 100. The score of each indicator is decided by a threshold approach, where the evaluated politician receives full points if the threshold is achieved (e.g. prepare at least two monitoring reports). Figure 1 provides an overview of the scorecard structure.

To assess the level of fulfillment of the councilors’ duties, qualitative data is collected through, inter alia, documentary reviews, interviews, personal diaries and field visits (observations and photos). When collected, the data is cleaned and reviewed centrally at ACODE to ensure accuracy and completeness and entered into Atlas.ti and EpiData for analysis (Grossman and Michelitch, 2018).

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\(^1\) The examples inside parentheses are parameters on which ACODE assess the fulfillment of the legally defined duties, as described in ACODE’s “Researcher’s manual” (which is not a public document).
Figure 1: Structure of scorecard

<table>
<thead>
<tr>
<th>PARAMETER/INDICATOR</th>
<th>Actual Score</th>
<th>Maximum Score</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. LEGISLATIVE ROLE</td>
<td></td>
<td>25</td>
</tr>
<tr>
<td>i) Participation in plenary sessions</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>ii) Participation in Committees</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>iii) Moved motions in Council</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>iii) Provided special skills/knowledge to the Council or committees</td>
<td>4</td>
<td></td>
</tr>
<tr>
<td>2. CONTACT WITH ELECTORATE</td>
<td></td>
<td>20</td>
</tr>
<tr>
<td>i) Meeting with Electorate</td>
<td>11</td>
<td></td>
</tr>
<tr>
<td>i) Office of coordination centre in the constituency</td>
<td>9</td>
<td></td>
</tr>
<tr>
<td>3. PARTICIPATION IN LOWER LOCAL GOVERNMENT</td>
<td></td>
<td>10</td>
</tr>
<tr>
<td>i) Attendance in sub-county Council sessions</td>
<td>10</td>
<td></td>
</tr>
<tr>
<td>4. MONITORING SERVICE DELIVERY ON NATIONAL PRIORITY PROGRAMMES AREAS</td>
<td></td>
<td>45</td>
</tr>
<tr>
<td>i) Monitoring of Health Service delivery units</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>ii) Monitoring Agricultural Projects</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>iii) Monitoring Education facilities</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>iv) Monitoring Road projects</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>v) Monitoring Water facilities</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>vi) Monitoring Functional Adult Literacy programmes</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>vii) Monitoring Environment and natural resources</td>
<td>5</td>
<td></td>
</tr>
</tbody>
</table>

Source: Grossman and Michelitch (2018)

Intense Dissemination (treatment)

The scorecards have been diffused among citizens at two occasions between the 2011 and 2016 elections through sets of community meetings on parish level (354 meetings with a total attendance of 12,939 in 2013 and 339 meetings with a total attendance of 14,520 in 2014). This is by ACODE referred to as the Intense Dissemination (ID) treatment. The ID treatment was put in place in collaboration with a team of external researchers in order to create an experimental setting for quantitatively evaluating the program. The description of the treatment in this subsection is based on the paper by Grossman and Michelitch (2018) who were a part of this team.

The ID treatment consists of several components, all aiming to make the scorecards common knowledge in the implementation sub-counties. As a first step to inform citizens about the scorecards, ACODE sent out invites to community meetings, targeting lower-tier government officials (at village and sub-county level), religious leaders, public service providers (e.g. teachers and health workers) and leaders within civil society. The meetings were open to the public and district councilors were invited to get a chance to comment on their scores as well
as on the initiative as a whole. In these meetings, on average 40 community members participated. Attendees were there educated on the scorecard initiative, its goals, the legally stated duties of councilors and the legally defined public service delivery standards. Following the meetings, attendees were tasked with disseminating the information to the communities. For this purpose, they were handed fliers, posters and calendars with key information of the scorecard to distribute and hang up in prominent places. In the disseminated scorecards, the councilors’ scores were benchmarked against those of other politicians in the LGCSCI implementation districts. In addition, the attendees received periodic text-messages reinforcing the key information delivered at the meeting and encouraging them to sign up others for receiving text messages with the scores of their councilors.

Facilitators employed by ACODE were tasked with delivering the information at the dissemination meetings. To track their compliance, in terms of delivering their specifically assigned meeting content, the research team employed enumerators to all meetings. A short poll was also conducted on random attendees to estimate the comprehension and retention of the information. Among the sampled attendees (n=1766), 56% recalled the councilors’ score, 98% could name at least one public service delivery standard and each individual recalled on average 2.75 councilor duties. The attendees’ compliance regarding the information dissemination tasks was however not investigated, which is why we cannot be entirely sure of the extent to which the scorecards actually became common knowledge in the communities, i.e., to what extent the individuals in the treatment areas actually were treated.

**Evaluated positions**

The district council positions evaluated in the scorecards are the regular councilors (directly elected representatives of each sub-county within a district) and woman councilors (elected by the women, representing 1-3 sub-counties). While there are also councilors representing the youth and people with disabilities, these positions were not included in the scorecard since they have a whole district as a constituency (Grossman and Michelitch, 2018). That woman councilors are the units of randomization should make treatment spillovers implausible since the regular councilors in the constituencies of the treated woman councilors also were treated. The untreated sub-counties hence received no information on any of their elected councilors and should be unlikely to react to information about councilors from other sub-counties, which it is also unlikely that they received.
Weak Dissemination

In the between-elections period, 2011-2015, scorecards were annually presented to incumbents, district officials and party representatives in dissemination events held at the district headquarters where also local stakeholders, such as journalists, civil society organizations and traditional leaders, were invited. Following Grossman and Michelitch (2018) I refer to this as “weak dissemination” (WD), since information about the scorecards only might trickle down to “ordinary” citizens who did not participate in these events through, for instance, word of mouth or media reporting. Since the WD occurred at district level in all ACODE districts, all sub-counties assigned to the ID treatment were also subject to the WD.

SMS treatment

Another treatment implemented parallel to the ID was a SMS platform where citizens, for free or for a small cost (depending on the operator), could send text messages directly to their district councilors informing them about issues related to public services. Politicians were informed about and trained to use the SMS platform at community meetings while the public received information through radio adverts. Similarly to the ID treatment, the SMS treatment was randomly assigned to sub-counties within the ACODE districts (ibid. 2018). This means that sub-counties within ACODE district either received the ID treatment, the SMS treatment or both.

3. Theoretical framework and empirical literature

While the relationship between information and forms of participation other than voting is under-theorized, there exists a large pool of empirical literature investigating this relationship (Fox, 2015; Gaventa and McGee, 2013). In this section, I first review the previous literature and discuss how it relates to the thesis. I then provide a theoretical framework of concepts relevant to understand the intervention studied and the mechanisms linking information to behavioral changes among citizens. Finally, I formulate testable hypotheses.

3.1 Empirical literature

This thesis aims to study the premise that information will leverage accountability through increased pressure from citizens. That is also the central premise of the program studied, LGCSCI (Bainomugisha et al., 2017). If citizens do not act as a result of the information, the
accountability relationship between citizens and politicians will not be strengthened. It is thus useful to look to some of the empirical work examining the information-accountability link, what lessons can be drawn from previous research and how it relates to this thesis.

One of the most influential studies demonstrating the potential for information to strengthen accountability is Björkman and Svensson’s (2009) field experiment in Uganda. Studying a community scorecard initiative targeting the health sector the authors find that treated communities saw substantial improvements in child mortality, immunization rates, service utilization, waiting times and absenteeism. This effect was achieved not through increased health sector spending but through disseminating information about the communities’ collective views of the performance in the different components of health services.

Like Björkman and Svensson, many studies that investigate the effects of information overlook the behavioral channels and focus directly on public service delivery outcomes. Besley and Burgess (2002) find that areas where media access is high see greater government responsiveness to calamities and drops in food production. Reinikka and Svensson (2005) examine a newspaper campaign informing citizens about officials’ handling of public grant, finding that areas with higher newspaper penetration saw both higher enrollment rates and test scores. In both these studies it remains unclear whether the improved outcomes were the result of increased pressure from citizens or if it was the awareness of government officials and service providers that their actions were being monitored.

The literature examining behavioral responses to information is less optimistic, which indicates that bottom-up pressure might not have been the main factor leading to the improved outcomes in the aforementioned papers. The two areas in which behavioral responses to political information have been most widely studied are perhaps education and voting.

Within the field of education, scholars have investigated the potential of information to trigger actions to improve schooling among students, teachers and parents. The outcomes investigated include teacher and student absenteeism (Barr et al., 2012); literacy and private investments into children’s education (Keefer and Khemani, 2011); parental involvement and school performance (Banerjee et al., 2010); student test scores and enrollment (Andrabi et al., 2017); and a range of public and private actions taken by parents (Lieberman et al., 2014). Among these papers, Andrabi et al. (2017) is the sole paper to find unambiguously positive effects of information.
A large part of the information-voting literature has studied how information affects preferences for candidates, such as left-wing or right-wing and incumbent or challenger. The most relevant outcome for this thesis is however voter turnout since it is a question of participating or not participating, as is attending community meetings and reporting problems. In a quasi-experiment in Benin, Fujiwara and Wantchekon (2013) find that groups that in town hall meetings discussed politicians’ policy platforms of broad-based public provision saw no difference in voting turnout relative to groups receiving standard clientelist campaign messages. In another experiment, closely related to the one in this thesis, citizens in Indian slums received report cards on the performance, wealth, education and criminal record of the incumbent and the two major challengers in a jurisdiction prior to an election. The findings showed that voter turnout was 3.5% higher in slums that received the information compared to slums receiving no such information (Banerjee et al., 2011).

Very few papers have studied the effects of information on non-institutionalized forms of public participation. This is a surprise considering the great magnitude of donor support to programs focused on mitigating informational constraints believed to hamper citizens’ participation in developing countries (Lieberman et al., 2014). Within the scarce literature studying effects of information on public participation outside of voting and not directed towards improving a specific public service, Krishna (2006) creates an index of individuals’ participation, which includes participating in meetings, campaigning and voting, based on survey answers. He finds that individuals that have greater access to information (measured as the number of information sources regularly accessed by a respondent, such as newspapers, radio, neighbors and leaders) see higher levels of participation. In another paper, Gottlieb (2016) finds that civic education meetings providing citizens with information on performance standards of local governments led to more citizen challenges of politicians in town hall meetings.

Out of the literature discussed so far, these two papers are perhaps of the greatest relevance for this thesis, but differ nevertheless. The Krishna paper does not look at any particular information content but rather general access to information. The participation outcomes are also general and not tied to a specific issue, such as service provision. This is also true for the paper by Gottlieb, which only looks at general challenges to local leaders. She also limits her investigation subjects to only including meeting participants and do not ask whether the information had any impact on meeting attendance. This thesis, in contrast, looks at participation, in the form of attending and speaking at meeting and reporting problems,
directly tied to public service delivery issues and asks whether such participation is triggered by information specific to the performance of locally elected leaders.

There are many explanations to why information about politicians might not have the desired effect of stimulating positive action. It might be that citizens are indifferent to the performance of politicians and are instead mainly influenced by ethnic politics or clientelistic arrangements (Humphreys and Weinstein, 2012). Increased transparency could also lead to corruption or shirking being hidden in ways that citizens cannot detect, reducing or offsetting the positive impact of information (Olken, 2009). Scholars studying the effects of corruption information has found that the information, if negative, can lead to adverse behavioral changes, such as staying away from the voting booth (Chong et al., 2010) or engage in corrupt behavior (Corbacho et al., 2016). Similarly, better informed citizens have been shown to be more reluctant to participate in authoritarian settings since it may seem futile or serve to legitimize the regime (Croke et al., 2016).

In summary, by looking at some of the areas in which behavioral responses to information have most widely been studied, this section has shown that the effects of providing citizens with information are uncertain and can go in either direction depending on the type of information and the outcome considered. While I have failed to find any literature examining the specific outcomes used in this thesis, there are nevertheless lessons to be drawn from the literature. Authors have emphasized the cost for citizens to use their voice as an important determinant of participation (Banerjee et al., 2011; Krishna, 2006; Paul, 1992). This could be a potential explanation to why the voter turnout-literature appears to be more optimistic than the literature on education. The effort associated with outcomes such as higher test scores or parental involvement in children’s learning and the school system should be higher than casting a vote on a single occasion. Similarly attending meetings is more time-consuming than voting and the opportunity costs of participating can be high, especially for poor people. However, as we shall see, the cost associated with participation is only one of many possible explanations to why an information campaign succeeds or not.

3.2 Theoretical framework

3.2.1 Accountability

To strengthen the accountability relationship between citizens and local politicians is the main aim of the intervention studied. This is because accountability widely is seen as a pivotal
component in good governance as well as in effective delivery of public services (Besley and Ghatak, 2003). To understand what accountability means, we can imagine a scenario where there are two actors: A is the locally elected politicians and B is the citizens. According to Fearon (1999), there are two conditions that need to be met for accountability to be in place. First, A is required to act on behalf of B in some sense. Second, B is empowered to sanction or reward A for her activities or performance in this capacity.

A useful distinction when talking about accountability is that between horizontal and vertical accountability. Horizontal accountability mechanisms are located within the state apparatus and requires government officials and agencies to report to other officials and agencies (Ackerman, 2004). Examples of such agencies include ombudsmen, corruption control agencies and administrative courts. Vertical accountability, in contrast, refers to the ways in which citizens and civil society organizations can hold governments accountable. Freedom of press, speech and association provides channels through which citizens can articulate their voice and strengthens vertical accountability (O'Donnell, 1998).

One of the most applauded vertical accountability mechanisms is free and fair elections where citizens elect representatives and, based on their performance, reward or sanction them at the ballot box. Recently, however, researchers have begun to acknowledge the deficiencies in elections as an accountability instrument (Ackerman, 2004; Gaventa and McGee, 2013). Malena et al. (2004), for instance, argue that elections are a blunt accountability instrument since citizens there cannot fully express their preferences and views, nor do they allow citizens to hold public officials accountable between elections or for specific decisions and behaviors.

This has led a new type of accountability initiatives to emerge, collectively termed social accountability initiatives, which do not suffer from the above-mentioned limitations of elections. These initiatives, of which the initiative evaluated in this thesis is an example, are also referred to as citizen led, bottom-up or demand-side accountability initiatives and encompass a broad range of actions and mechanisms made available for citizens to interface with actors, such as governments, in ways that are social rather than political, institutional or bureaucratic (Gaventa and McGee, 2013). The idea is to promote participation beyond voting among citizens and societal actors in order to hold governments and other power-holders within the state accountable. According to Brinkerhoff and Wetterberg (2016), social accountability initiatives have three instrumental aims: increasing the effectiveness of service
delivery, improving the quality of governance and democracy and increasing citizen empowerment. Out of these interlinked instrumental aims, the initiative studied in this thesis primarily aims to increase citizen empowerment and, in particular, empower citizens to participate and demand improved public services. In summary, social accountability initiatives serve to both compliment and enforce horizontal and vertical accountability mechanisms (Malena et al., 2004).

3.2.2 Conditions under which information affects citizen behavior

For information to have any effect on citizen behavior, certain conditions regarding the information, the citizens and the political environment need to be fulfilled. Below I propose conditions upon which information provided by a CSO to induce behavioral change among citizens is contingent, as suggested in the literature.

First, the information needs to be reliable, which means that the issuer or disseminator of information must be credible and legitimate (O’Meally, 2013). Credible and rich information can incentivize individuals and communities to use their own resources to call for change (Barr et al., 2012). If citizens do not trust the information to be accurate they might not see a behavioral change or choose inactivity (De Figueiredo et al., 2011).

Second, citizens need to understand the content of the information. If citizens are unable to interpret the available information they are unlikely to take action (Banerjee et al., 2011). To facilitate citizens’ understanding, the information must be clear. Fox (2007) makes a distinction between clear and opaque transparency. Programs that report reliable information about institutional performance fall into the former category and allow the concerned actors, such as the public, to act on the information and pursue strategies of constructive change. Opaque transparency, in contrast, implies to provide information distorted from the practical reality. Training and education can help citizens internalize the information (Krishna, 2006).

Third, the attitudes and prior beliefs of citizens play an important role in determining the impact of the information. Citizens must be dissatisfied with the current state of affairs and want an improvement (Kosack and Fung, 2014). However, information describing the poor performance political performance could also lead to a reluctance to participate among citizens if, for instance, the expected benefits of participating is reduced (Banerjee et al., 2011). Furthermore, if information about local government performance merely confirms what an individual already thinks, any behavioral change is unlikely. It is thus important to raise citizens’ expectations about what is an acceptable performance. The way to do this, as
suggested by Gottlieb (2016), is to make sure that citizens have accurate reference points regarding the performance of governments and politicians to use as evaluative criteria. These reference points could either be a general performance standard or a benchmark against the performance of other politicians.

Fourth, the political environment could be a serious impediment to citizens’ voice. An enabling environment needs to reduce fear of reprisals for citizen action. When possible, channels where citizens anonymously can use their voice should be promoted (Fox, 2015).

The political environment needs to allow citizens to easily identify who is responsible for the service delivery performance, whether it is frontline service providers, politicians or bureaucrats. In developing countries with multiple levels of government, there can also be uncertainties regarding which level is responsible for the provision of a specific service (Chong et al., 2010; Gottlieb, 2016). These issues are especially prevalent in public service environments characterized by what Booth (2010) refers to as institutional incoherence. Common features of these environments are ill-defined mandates, overlapping jurisdictions, the pursuit of impractical policies and distorted incentives among actors within implementing organizations.

Fifth, citizens need to be able to and willing to act on the information. Without sufficient skills and resources citizens are unlikely to have capacity or be motivated to act (Brinkerhoff and Wetterberg, 2016). Prior knowledge of the political system, what concrete impact can be sought and what channels can be used to effectively demand change, facilitates taking action. If the information has a clear connection to citizens’ well-being a reaction is more likely. Banerjee et al. (2011), for instance, show that voters react to legislators’ attendance record in oversight committees but not in the legislature and that citizens are concerned with public goods spending in areas where they live but not in general. Similarly, a person that spends a substantial portion of her time and income on a public service is more likely to voice potential concerns (Paul, 1992). That is, it is important that citizens care about the information (Kosack and Fung, 2014). Seeking to influence government decisions requires of citizens both a belief in their own capacity and in a government responsive to pressure (McLeod et al., 1999).

Sixth, to facilitate collective action, citizens need to expect others to act towards a common good (Ostrom, 1998). Therefore, attention to citizens’ incentives, both for individual or collective action, is important (Booth, 2012). Coordination among citizens becomes more likely when citizens know that others are receiving the information as well (Lieberman et al.,
2014) and when groups, such as a constituency, share interests and moral obligations (Tsai, 2007). Participation in meetings can also facilitate learning about politicians’ performance through common understandings of benchmarks (Fujiwara and Wantchekon, 2013) and what the expectations and targets are (Barr et al., 2012).

While these conditions for information to trigger behavioral responses from citizens are not exhaustive, they nevertheless provide some insights into the mechanisms that can link information to behavioral changes among citizens.

### 3.3 Hypotheses

I first expect the scorecard dissemination to have positive effects on citizen participation. While this is far from obvious, I argue that it is likely that many of the conditions listed in the previous subsection are fulfilled in the context studied. This factor, together with the positive findings in the papers by Gottlieb (2016) and Krishna (2006), which I previously argued are the papers most related to this thesis, lead me to expect positive effects. I discuss below each condition and, based on the context, offer conjectures in regards to whether a condition is likely to be fulfilled or unfulfilled.

First, the disseminator of information (ACODE) needs to be perceived as legitimate by citizens. ACODE’s assessment of politicians is intended to be non-political, fair and objective (Bainomugisha et al., 2017), which should contribute to ACODE being seen as legitimate and credible.

Second, citizens need to understand the information. While I am unable to assess this for the majority of the treated who solely were exposed to the scorecards, the attendees of the ID meetings appear to have understood the information to some extent. Following the meetings, 98% of the attendees could mention at least one public service delivery standard and 56% recalled their councilor’s score.

Third, citizens’ attitudes toward local government might not favor participation. At baseline, 60% of respondents in the SLRC survey believed that local government does not care about their opinions. If citizens do not believe that a government is responsive to their pressure, a reaction is less likely (McLeod et al., 1999). For the information to have any effect, it needs to make citizens update their prior beliefs. This appears to be the case. Grossman and Michelitch (2018) find in a baseline survey that only 9% of respondents had heard about the scorecard initiative prior to the ID treatment. They also find that respondents’ assessments of
councilors’ performance along the four legally defined duties were uncorrelated with the actual scorecard scores. Furthermore, the information needs to be clear on what is an acceptable performance. The disseminated scorecards not only described the scores of citizens’ own councilors, but also the scores of councilors representing other sub-counties. These benchmarks should make it clear to citizens what they can expect from their councilors.

Fourth, the political environment needs to favor citizens’ use of voice. As observed by Titeca (2005), citizens in Uganda do not appear to be afraid to confront government officials at meetings. If a government official would blatantly lie about some issue, the meeting attendees often do not hesitate to speak up and confront the official. Furthermore, it needs to be clear to citizens what level of government is responsible for delivery of a service. The scorecards describe the services that the councilors are responsible for monitoring and it should hence be clear to citizens that they can hold the councilors accountable for the performance in that respect.

Fifth, citizens need to be able to and willing to act on the information. The low level of development in Uganda is an impediment to this condition. The opportunity costs of participating can be discouraging, especially for poor people (Steiner, 2007). The dataset used shows that 68% of respondent households are unable to meet the household needs and sometimes or often need to rely on others for help. These individuals might be reluctant to participate even after receiving the new information.

It has been suggested that citizens in low-income settings might be relatively indifferent to the performance of politicians and instead have greater preferences for clientelistic arrangements or private transfers (Humphreys and Weinstein, 2012; Pande, 2011). If this is the case, citizens might not observe an increased motivation to act as a result of the ID treatment, given that it does not compromise the set arrangements. Grossman and Michelitch (2018) report in the baseline survey that 41% of respondents believed paying personal handouts is a legal responsibility of councilors, which indicates that clientelistic arrangements do occur. However, I cannot assess the importance citizens give to such arrangements relative to the performance indicators in the scorecard. Furthermore, citizens’ ability to act is facilitated by some prior knowledge of the political system and channels through which they can influence politicians. The dataset shows that 53% of respondents were unaware of any official channels through which they can report a service delivery problem and 51% were unaware of any community meeting having been held.
Sixth, the unawareness of influence channels and meetings held can also be an impediment to collective action. If citizens that are aware of how to take action believe that few others are informed of how to act, the transaction costs for collective action might appear too high for the informed citizens (Buntaine et al., 2018).

As we have seen, it is far from clear that all of the conditions are fulfilled and, in fact, it appears rather unlikely. However, several conditions seem likely to have been fulfilled. The information should be new to citizens and update their prior beliefs, meeting participants seem to have understood the information and the political environment is not characterized by fear among citizens to use their voice. In addition, the scorecard dissemination itself could assist in the fulfillment of other conditions by clarifying the responsibilities of district councilors; moving citizens’ evaluative criteria of councilors from paying handouts and private transfers to performing their legal duties; and creating a buzz around councilors’ performance which could facilitate collective action. The somewhat large degree to which the conditions appears to be fulfilled, together with the positive information effects found in the previous literature most relevant for this thesis (Gottlieb, 2016; Krishna, 2006), leads me to hypothesize that the scorecard dissemination should have a positive effects on the six participation outcomes.

**Hypothesis 1:** *The scorecard dissemination will induce more citizens to attend meetings, speak at meetings and report service delivery problems.*

In addition to this main hypothesis, I make several testable predictions regarding heterogeneity in impact for individuals with different attributes, which I base on previously made arguments in the thesis. First, citizens need to be dissatisfied with service delivery quality and local government service delivery efforts in order to react to information (Kosack and Fung, 2014). Second, a belief that local government is responsive to the voice of citizens should facilitate citizen pressure on local governments to improve performance (McLeod et al., 1999). Third, information might have lesser impact on poor citizens in Uganda due to higher opportunity costs of participating (Steiner, 2007) and a perception that it is the elite group and not the poor that should participate (Steiner, 2008). Fourth, educated citizens should be more likely to participate due to their larger capacity to internalize the information as well as their better knowledge of the political system and how to act. These arguments are summarized my second hypothesis:

**Hypothesis 2:** *The scorecard dissemination will have a larger effect for individuals that:*
- believe that local government is responsive to their opinions
- are dissatisfied with local government decisions
- are dissatisfied with the quality of public services
- have higher living standards
- are better educated

4. Data

The main data used in this study comes from the Secure Livelihoods Research Consortium (SLRC) Panel Survey - a multi-year survey carried out in the northern Ugandan sub-regions Acholi and Lango in 2013 (January-February) and 2015 (January-February). The first round of scorecard dissemination occurred between June and August 2013 and the second round between March and July 2014, meaning that the survey rounds took place roughly 6 months before and after the treatment period. The SLRC dataset contains modules on livelihood sources, food security, security, shocks, basic services, social protection, livelihood services and governance. For the purpose of this thesis, the wide range of questions included in the survey can be seen as an advantage since this should reduce the risk of priming the respondents.

At baseline 1853 individuals were surveyed, out of which 1553 remained in the second wave, meaning an attrition rate of 16%. The sampling method, proportional to size systematic sampling (PPSSys), means that larger sub-counties have a higher likelihood of being selected in order to equalize each respondent’s probability of selection. To address these two concerns of (non-random) attrition and unequal probabilities of selection, the observations in both waves are assigned design weights. For a more detailed description of the survey methodology, see Marshak et al. (2017).

In addition, I collect data on the ID and SMS treatments from Grossman and Michelitch (2018). The ID and SMS treatments were randomly assigned to sub-counties within the 20 districts where ACODE is active.

4.1 Measurements of active citizenship

From the SLRC dataset I construct six participation measurements, out of which three are more general and three are more strongly connected to local governments. I begin by
describing the general participation outcomes and then explain how these differ from the local government (LG) participation outcomes.

The variables used to capture citizen participation and demand for improved public services are *attended meeting, spoke at meeting* and *reported problem*. Out of these, the first two are more related to collective action and the third to individual action. Depending on the context, individual and collective action can be perceived as more or less appealing and it is thus useful to investigate these separately (Dawes, 1980).

*Attended meeting* is a dummy variable taking the value 1 if an individual answered yes to having attended a community meeting regarding any public service in the last 12 months. This question was asked for each of the public services: health, education, water, livelihood assistance, social protection, security and other services. I code this variable as taking the value 1 if a respondent answers “yes” to having attended a meeting regarding any of those 7 public services. Hence, the variable captures whether an individual participates or not, but not the level of activity of an individual. Since it has been suggested that the level of activity among citizens vary largely between low participators and high participators, with the former being generally inactive and the latter participating in a number of different political activities (Krishna, 2006; McLeod et al., 1999), and that effective participation requires collective action by a substantial number (Booth, 2012; Ostrom, 1998), I argue that focusing on the transition from being a non-participant to a participant perhaps is more meaningful than looking at the overall level of activity of individuals when measuring active citizenship in this context. All dependent variables are hence coded in the same fashion.

*Spoke at meeting* is a dummy variable indicating whether the respondent has spoken at a community meeting (regarding any of the public services mentioned above) in the last 12 months.

*Reported problem* is constructed in the same way as the previous two variables, taking the value 1 if the respondent answers yes to have reported a problem regarding any of the 7 public services in the last 12 months.

The LG participation outcomes differ from the above described general participation outcomes in that, for an individual to be consider a participant, he/she is required to answer yes to a follow-up question, given that the answer was yes to having attended a community meeting or reported a service delivery problem. A respondent that answered yes to having
attended a community meeting was subsequently asked who called the meeting, whether it was an NGO, a local clan leader, a government official, a health worker, a religious leader, a local extension worker, a community group or a community security group. The most straightforward response to the ID treatment would be to attend or speak at a meeting called by a government official. Therefore, the new variables for attending a community meeting and speaking at a community meeting are coded as 1 if an individual took each action (spoke and attended) for any public service at a community meeting called by a government official. Since it is unlikely that central government officials call community meetings in northern Uganda, the government official in question should pertain to a local government. Hence, to avoid confusion, I refer to meetings called by government officials as local government meetings.

For the variable *reported problem*, the follow-up question was to whom the problem was reported, with the options: the local council, the community, an international agency, a local non-governmental organization (NGO), a religious institution, a private provider or other. The alternative version of *reported problem* is coded as 1 if an individual reported a problem to a local council.

Since the participation measurements are meant to capture actions that citizens take to exert pressure on local governments to improve service delivery, the local government participation outcomes can appear as more straightforward. However, there are other ways in which citizens can exert pressure than taking actions directly directed towards local government. Actors such as religious leaders, NGOs or clan leaders can serve as a communication link between citizens and local governments, amplifying the voices of citizens. As pointed out by Titeca (2005), NGOs are a prominent link between citizens and local governments in Uganda. They hold meetings where citizens can express their grievances to local government officials. NGOs also visit the communities, gather citizens’ views on issues and compile reports to be presented to local government officials. One could also argue that, since local governments are responsible for the majority of public services in Uganda, any meeting attended where public services are discussed can be a way of indirectly pressuring local government to improve public services by, for instance, facilitating collective action and mobilization for future attempts to influence local government decisions. However, it is possible that the role of frontline service providers rather than the role of local governments is discussed at these meetings. Perhaps citizens see deficiencies in the local health centre as being the fault of the staff and make no connection to local government efforts. Therefore, I argue that it is useful
to separate general participation outcomes from local government participation outcomes. The former outcomes are a black box in the sense that I cannot be entirely sure if the participation they capture is related to local government, whereas the latter more certainly is related. It is hence likely that the treatment effect is larger for the local government outcomes since they provide a clearer link to the ID treatment. Table 1 summarizes the outcome variables and provides the count of untreated and treated observations for each variable. The table shows that roughly half of the individuals that attended a community meeting and spoke at a community meeting did so at a local government meeting. 905 out of the 1405 individuals that reported a service delivery problem did so to the local council.

<table>
<thead>
<tr>
<th>Participation outcome</th>
<th>Count of untreated observations</th>
<th>Count of treated observations</th>
<th>Count of total observations</th>
</tr>
</thead>
<tbody>
<tr>
<td>Attended meeting</td>
<td>2004</td>
<td>1405</td>
<td>3409</td>
</tr>
<tr>
<td>Spoke at meeting</td>
<td>2362</td>
<td>1047</td>
<td>3409</td>
</tr>
<tr>
<td>Reported problem</td>
<td>2039</td>
<td>1370</td>
<td>3409</td>
</tr>
<tr>
<td>Attended LG meeting</td>
<td>2686</td>
<td>723</td>
<td>3409</td>
</tr>
<tr>
<td>Spoke at LG meeting</td>
<td>2895</td>
<td>514</td>
<td>3409</td>
</tr>
<tr>
<td>Reported problem to LG</td>
<td>2504</td>
<td>905</td>
<td>3409</td>
</tr>
</tbody>
</table>

4.2 Control variables

The covariates include individual characteristics (age, sex), education (highest level of education completed), indicators of living situation (living standard, food security), connectivity (owns phone, owns radio), health (household member experienced long term health problems) and social capital (able to borrow money from friends/family). NRM councilor is a sub-county characteristic indicating whether a sub-county is represented by a member of the National Resistance Movement at the district council. These variables are described in Appendix A1.

SMS treatment is a binary variable indicating whether a sub-county was assigned an SMS platform where citizens can make direct contact with their elected local politicians (see section 2.2). Out of the outcomes investigated, this treatment is only believed to influence whether a citizen has reported a problem, since the purpose of the SMS platform was to allow citizens to express grievances regarding public services to district councilors.
Weak dissemination is a variable taking the value 1 if an individual belongs to a district where ACODE-activities are conducted. Recall that ACODE, in all districts where present, holds annual scorecard dissemination events where politicians and other stakeholders are invited. Since the scorecard information through these events might trickle down to citizens, the effect of the ID treatment might be overestimated if the variable is not controlled for.

Both variables attended meeting and spoke at meeting are contingent on a community meeting being held in the last 12 months. Therefore community meeting held is a considered control variable. By similar reasoning, to have experienced a service delivery problem is necessary for having reported a service delivery problem. Both variables experienced a problem and community meeting held are perception-based and might not accurately reflect the true state, but they should nevertheless be considered as controls.

5. Research design and empirical strategy

This thesis is predominantly interested in the Intense Dissemination (ID) component of the Local Government Councils Scorecard Initiative (LGCSCI) where scorecard information is disseminated to citizens. The ID treatment is randomly assigned to sub-counties within the districts in which the LGCSCI is implemented by ACODE. The reason why focus is on the ID treatment and not on the Weak Dissemination is the uncertainty regarding the extent to which the annual dissemination events garnered attention from the media and, in case it did, to what degree citizens took part in the media reporting. Given the Ugandan context with low levels of connectivity among citizens and a relatively scarce number of media outlets, it appears unlikely that these events should have any considerable effects on citizens’ awareness of the scorecards. The ID treatment, in contrast, was designed make citizens in the treated sub-counties aware of the program and the score of their local councilors, which makes it a more suitable treatment to investigate given the purpose of the thesis.

An issue with estimating the causal effect of the program is that individuals are not entirely randomly selected into treatment and control groups. ACODE is present in 20 Ugandan districts, out of which 4 are included in the SLRC survey (see Figure 2). Since the ID treatment was randomly assigned to sub-counties within the ACODE districts, only a part of the sub-counties in the sample was eligible for ID treatment. The question is then if the districts chosen for ACODE-activities are different in characteristics from districts without
ACODE presence. This could plausibly be the case since ACODE needs the approval and cooperation of the politicians within a district to conduct their activities, which suggests that the political environment, as well as other characteristics, could be different in non-ACODE districts. Should the political environment in the districts chosen by ACODE particularly favor participation, the treatment effect is likely to be biased upward and treatment assignment cannot be considered random.

When one does not have the benefit of a setting where treatment is randomly assigned there are, however, empirical methods available to ensure that treatment is “as good as random”. Two such methods frequently used in the impact evaluation literature are difference-in-difference (DID) and propensity score matching (PSM). Both these methods can be used to estimate causal effects when one has a group of treated individuals and a group of untreated individuals observed at two time periods. Since these methods rely on different assumptions and have different advantages, they can be seen as compliments. Hence, to assess the robustness of the findings, I estimate results using both DID and PSM. In addition, I estimate heterogeneous treatment effects using the DID method as base.

![Figure 2. Map of Uganda indicating ACODE districts and districts in the SLRC dataset](image-url)
5.1 Difference-in-difference

The difference-in-difference (DID) method compares outcomes over time between a group subject to treatment and an untreated group. The first difference, assessing differences over time in the two groups, accounts for time-constant effects since each group here is compared with itself. This leaves time-varying factors the only remaining source of bias. Given that time-varying factors affect the treatment and control group in the same way, comparing the first differences between the two groups (second differences) accounts for these time-varying factors. Hence, by subtracting the second difference from the first difference we have, in theory, removed time-constant and time-variant effects (Gertler et al., 2016). The intuition behind the DID estimation technique is illustrated in Table 2, using the data of this thesis as an example. Note that design weights are not applied which gives the estimate little credibility as the actual treatment effect.

<table>
<thead>
<tr>
<th>Group</th>
<th>Mean of variable attended community meeting</th>
<th>Change After-Before</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>After=2015</td>
<td>Before=2013</td>
</tr>
<tr>
<td>Treated</td>
<td>0.4260</td>
<td>0.3802</td>
</tr>
<tr>
<td>Control</td>
<td>0.4231</td>
<td>0.4063</td>
</tr>
<tr>
<td></td>
<td>Difference-in-difference estimate:</td>
<td></td>
</tr>
</tbody>
</table>

The first difference for the treated group is calculated by subtracting the average of the variable attended community meeting in 2013 from the average of the variable in 2015 (0.2460-0.3802=0.0458). After calculating the first difference for the control group in the same way, the second difference is calculated by subtracting the first difference of the control group (0.0168) from the first difference of the treated group (0.0458). This gives the difference-in-difference estimate 0.0290.

If the differences between the treatment and the control group vary over time, the DID fails to account for this. Therefore, the method requires the assumption that treatment and control groups would follow the same trends (conditional or unconditional on covariates) in absence of treatment. This is the parallel trends assumption and it is the key identifying assumption in DID.

There are both pros and cons for including covariates in a DID framework. If different trends among the treated and control group are driven by the characteristics of the two groups, relevant variables can be included to account for this. On the other hand, additional variables
make the common trends assumption more difficult to fulfill (Lechner, 2011). Important, however, is that these variables do not impact treatment in the second period which is referred to as the exogeneity assumption.

When data is available for treatment and control groups in two periods, DID can be estimated in a regression framework, which conveniently allows for adding more covariates and estimate standard errors (Angrist and Pischke, 2008):

\[ Y_{ist} = B_0 + B_1(T_s \times POST_t) + B_2T_s + B_3POST_t + \gamma X + \epsilon_{ist} \]

In this basic model, \( Y \) is a measure of active citizenship for individual, \( i \), in sub-county, \( s \), and at time, \( t \). \( T \) is a binary variable taking the value 1 for an individual in the treatment group and 0 for an individual in the control group. \( POST \) is a binary time variable taking the value 1 after the treatment and 0 before the treatment. \( B_1 \), the parameter of interest, is the interaction between \( POST \) and \( T \). This term is the difference-in-difference estimator with which I estimate the impact of the program. \( X \) is a vector of covariates and \( \epsilon \) is the residual. The vector \( X \) consists of individual characteristics, sub-county characteristics and alternative treatments in place.

**Parallel trends**

The identification of DID depends critically upon the parallel trends assumption, which therefore needs to be assessed. Conventionally, this is tested by examining pre-treatment years for trends but since I only have one pre-treatment year this cannot be done. However, some confidence for parallel trends can be provided by using a placebo outcome that is unaffected by the treatment instead of the actual outcome of interest, as suggested by Duflo (2002) and implemented in, for instance, Fowler (2013). If the parallel trends assumption holds, the treatment effect on this placebo outcome should be statistically indistinguishable from zero. For a placebo outcome I consider the variables \( NRM \) councilor, \( Sex \) and \( Age \). \( NRM \) councilor should not be affected by the treatment since the two survey rounds took place in the same between-elections period. The findings are reported in Table 3. In each test, the treatment effect is statistically insignificant which provides some support for the parallel trends assumption.
Table 3. Difference-in-Difference with placebo outcomes

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>Treatment effect</th>
<th>Standard error</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sex</td>
<td>-0.002</td>
<td>0.012</td>
<td>0.889</td>
</tr>
<tr>
<td>NRM councilor</td>
<td>0.015</td>
<td>0.012</td>
<td>0.184</td>
</tr>
<tr>
<td>Age</td>
<td>0.438</td>
<td>0.284</td>
<td>0.127</td>
</tr>
</tbody>
</table>

In each row, the treatment effect is the DID estimator regressed on the dependent variable at the left, with time fixed-effects and sub-county fixed-effects. NRM councilor and Sex are binary variables while Age is discrete. Standard errors are clustered by sub-county. Design weights are applied.

5.2 Propensity Score Matching

The idea of propensity score matching (PSM) is to compare individuals that are as similar as possible in characteristics but differ in treatment status. Imagine that there are two individuals, both identical in characteristics, with one receiving the treatment and one being a control. Comparing the two individuals’ outcomes would then show the causal effect of the treatment since being treated is the only way in which the two differ. Similarly, when there are two groups, one treated and one untreated, the average difference in outcome of these groups show the causal effect.

Matching individuals with the exact same characteristics one-to-one is referred to as exact matching and is believed to result in the most credible inference (Imbens, 2004). This, however, can result in dimensionality problems, i.e. difficulties to get matches when there are many covariates on which the matching is done (Caliendo and Kopeinig, 2008). The solution to this problem, proposed in Rosenbaum and Rubin (1983), is to instead match individuals on the probability of being treated, conditional on their observed characteristics unaffected by the treatment.

In PSM, the common trends assumption is not required, but instead it is assumed that all differences between treatment and control groups are identified, i.e., treatment only depends on observables. This is the Conditional Independence Assumption (CIA) which is the main identifying assumption in PSM. This contrasts the method to DID which, as previously described, allows for selection on time-constant unobservables.

A second important assumption in PSM is the “overlap assumption”, which requires participants and non-participants to have a positive probability of being selected for treatment. This assumption ensures that treatment observations and controls have a sizeable overlap in propensity scores (common support) so that suitable matches can be found.
The overlap assumption can be tested by visually inspecting the densities of propensity scores for the treated and the controls, before and after matching. This is done in Figure B1 to B3 in Appendix B and shows that the densities for both groups coincide well after matching. If the CIA and overlap assumption hold, we have what Rosenbaum and Rubin (1983) refers to as a situation of “strong ignorability” of treatment assignment.

In this thesis, the propensity scores (or probabilities of being treated) are estimated with a probit model, where relevant covariates are included. The parameter of interest in the PSM is the average treatment effect of the treated (ATT). ATT is defined at the difference in the expected outcomes of being treated and untreated for the individuals that actually participated in the treatment (Caliendo and Kopeinig, 2008). This can be expressed in terms of potential outcomes as:

\[
ATT = E(Y_1 - Y_0|T = 1) = E(Y_1|T = 1) - E(Y_0|T = 1).
\]

There are a variety of matching techniques to choose from when estimating propensity scores. I make use of the most common technique, Nearest Neighbor (NN) matching. When sample sizes are small, results can be especially sensitive to the choice of matching algorithm (Heckman et al., 1997). Therefore, to test the consistancy of the propensity score estimates, I use three different NN matching techniques.

In NN matching, treatment units are matched with the control units with the most similar propensity scores. This can be done with or without replacement. NN matching with replacement allows each non-participant to be matched several times with a non-participant, should it be the closest. The choice of whether to use replacement involves a trade-off between bias and variance (Caliendo and Kopeinig, 2008). While replacement will result in better matches it will also increase bias as a result of the reduction in unique nonparticipants. I estimate results using both NN with and without replacement.

In NN-matching with replacement the number of control units to be matched with each treated unit needs to be decided. To use more than one NN is referred to as “oversampling” and has the advantage that more information is being used. In the NN with replacement estimation, I allow for 5 controls units to be matched with each treated unit, which I refer to as NN5.

A risk in NN matching is to get poor matches. The nearest neighbor might not be so near, should there be large heterogeneity in the treatment and control groups. A solution is then to
impose a maximum allowed distance (caliper) in propensity score between the matched pair. As a third test, I impose a caliper of 0.003 to the NN5-model. A narrow caliper can improve the performance of PSM. However, if the caliper makes it difficult to find matches for many treated subjects this might result in selection bias and inefficiency due to a reduced sample size (Lunt, 2013). The caliper of 0.003 is found to not substantially reduce the matches.

In order for the treated and untreated groups to be comparable, their characteristics need to sufficiently balanced. That is, no statistically significant differences in covariate means between the two groups should exist (Heinrich et al., 2010). I test this before and after matching with t-tests, as suggested by Rosenbaum and Rubin (1985), which can be seen in Table B1, in Appendix B. While differences in means are expected before matching, there should be no such differences after matching. This can be seen are the cases for NN5 and NN with caliper but not for NN without replacement where a significant difference in means exist for the variable Weak Dissemination. Hence, less confidence should be given to this matching method than the other two.

5.3 Discussion of the main assumptions of DID and PSM in this context

Both DID and PSM require strong assumptions for identifications. The main assumption in PSM is the conditional independence assumption (CIA), which does not allow for selection on any unobservables. This assumption cannot be tested and one has to be justified by the available data (Caliendo and Kopeinig, 2008). Given the limited data at my hand, I cannot fully trust this assumption to be fulfilled. I argue that the DID model, which allows for selection on time-constant unobservables, is a better identification strategy in this setting. Also, unlike the PSM, I can for the DID provide some support for the main identifying assumption (parallel trends). I therefore consider DID my main model, while the PSM is used as a robustness check.

6. Results

In this section I present the results of the DID and PSM estimations. I also test for heterogeneous treatment effects building on the DID model. The first two subsections outline the results of the PSM and DID and are intended to answer Hypothesis 1: The scorecard dissemination will induce more citizens to attend meetings, speak at meetings and report
service delivery problems. The third section reports the results of the heterogeneous treatment effect assessment and is meant to answer my second hypothesis.

6.1 Difference-in-Difference

Difference-in-difference models are estimated for two versions of the outcome variables. In the first version, I investigate the effects of the ID treatment on the general participation outcomes and in the second version on the local government participation outcomes. The results of the first and second versions are reported in Table 4 and Table 5, respectively. For each of the 6 outcomes I estimate a model i) without covariate adjustment, ii) with the full set of covariates, and iii) excluding the conditioning variables (experienced a service delivery problem or community meeting held). The reasons for excluding the conditioning variables are that they 1) might violate the exogeneity assumption, i.e., be influenced by treatment in the second period and 2) are highly correlated with the respective outcome, which might be due to incorrect answers. On the other hand, an individual cannot attend a meeting if no meeting is held and is unlikely to report a service delivery problem if all public services are infallible. In all models, standard errors are clustered at the sub-county level allowing for serial correlation within, but not between, sub-counties. Design weights are applied to observations in both years.

From estimating the first version, using general participation outcomes, I find the treatment effect to be negative but insignificant in all models, which can be seen in Table 4. However, in the second version, reported in Table 5, I find some evidence of negative treatment effects. The treatment effect is negative and significant for the outcomes spoke at LG meeting (p-value=0.014) and attended LG meeting (p-value=0.028) when I include the full set of controls (column 2 and 5). In these models, we see that the inclusion of the conditioning variable reduces the standard errors of the DID coefficients, compared to column 3 and 6 where the conditioning variable is excluded, making the estimates significant at the 5% level.

Based on these key results for the thesis, I reject Hypothesis 1. Rather than having a positive effect on participation, the scorecard information appears to lead to disengagement or, at best, to maintaining the status quo.

Among the included covariates Age, Sex and Living Standard are significant predictors of participation in most models. Individuals that are older, male and better-off appear to be more likely to participate. Noteworthy is that the coefficient of SMS treatment is negative and significant in both versions (model 8 and 9, both tables). This is surprising considering that
the purpose of that treatment was to facilitate contact between politicians and citizens by establishing a platform for interaction through text messaging. Hence, if a citizen experiences a service delivery problem, it should require less effort to report it for citizens in sub-counties where the platform is implemented. It is possible that the radio adverts had limited outreach. Only 57% of respondents live in a household that owns a radio. Households that own a radio might also not have heard the radio advert or, if they have heard it, did not understand or believe in the idea of the SMS platform. The negative coefficient is, however, still difficult to explain.

Table 4. Difference-in-Difference results using general participation outcomes

<table>
<thead>
<tr>
<th></th>
<th>(1) Attended meeting</th>
<th>(2) Attended meeting</th>
<th>(3) Attended meeting</th>
<th>(4) Spoke at meeting</th>
<th>(5) Spoke at meeting</th>
<th>(6) Spoke at meeting</th>
<th>(7) Reported problem</th>
<th>(8) Reported problem</th>
<th>(9) Reported problem</th>
</tr>
</thead>
<tbody>
<tr>
<td>DiD</td>
<td>-0.042</td>
<td>-0.028</td>
<td>-0.044</td>
<td>-0.030</td>
<td>-0.019</td>
<td>-0.031</td>
<td>-0.068</td>
<td>-0.080</td>
<td>-0.067</td>
</tr>
<tr>
<td></td>
<td>(0.069)</td>
<td>(0.021)</td>
<td>(0.070)</td>
<td>(0.065)</td>
<td>(0.032)</td>
<td>(0.067)</td>
<td>(0.066)</td>
<td>(0.058)</td>
<td>(0.063)</td>
</tr>
<tr>
<td>Treated</td>
<td>0.014</td>
<td>0.024</td>
<td>0.044</td>
<td>0.075</td>
<td>0.082***</td>
<td>0.097</td>
<td>0.065</td>
<td>0.004</td>
<td>0.048</td>
</tr>
<tr>
<td></td>
<td>(0.051)</td>
<td>(0.015)</td>
<td>(0.068)</td>
<td>(0.058)</td>
<td>(0.033)</td>
<td>(0.074)</td>
<td>(0.081)</td>
<td>(0.062)</td>
<td>(0.082)</td>
</tr>
<tr>
<td>After</td>
<td>0.035***</td>
<td>0.001</td>
<td>0.009</td>
<td>0.002</td>
<td>-0.025**</td>
<td>-0.020</td>
<td>-0.052**</td>
<td>-0.050***</td>
<td>-0.036</td>
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<tr>
<td></td>
<td>(0.019)</td>
<td>(0.009)</td>
<td>(0.021)</td>
<td>(0.018)</td>
<td>(0.011)</td>
<td>(0.019)</td>
<td>(0.024)</td>
<td>(0.018)</td>
<td>(0.025)</td>
</tr>
<tr>
<td>Age</td>
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<td>0.002***</td>
<td>0.002***</td>
<td>0.003***</td>
<td>0.001***</td>
<td>0.000***</td>
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<tr>
<td></td>
<td>(0.000)</td>
<td>(0.001)</td>
<td>(0.000)</td>
<td>(0.001)</td>
<td>(0.001)</td>
<td>(0.001)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sex</td>
<td>-0.028**</td>
<td>-0.051**</td>
<td>-0.107***</td>
<td>-0.125***</td>
<td>-0.064***</td>
<td>-0.079***</td>
<td></td>
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<tr>
<td></td>
<td>(0.014)</td>
<td>(0.024)</td>
<td>(0.016)</td>
<td>(0.020)</td>
<td>(0.017)</td>
<td>(0.023)</td>
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<tr>
<td>Living standard</td>
<td>-0.022***</td>
<td>-0.064***</td>
<td>-0.022**</td>
<td>-0.054***</td>
<td>-0.028**</td>
<td>-0.002</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Education</td>
<td></td>
<td></td>
<td>0.002</td>
<td>0.000***</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td></td>
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<td>(0.003)</td>
<td>(0.002)</td>
<td>(0.002)</td>
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<td>Health</td>
<td>-0.017</td>
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<td></td>
<td>(0.012)</td>
<td>(0.020)</td>
<td>(0.021)</td>
<td>(0.023)</td>
<td>(0.017)</td>
<td>(0.019)</td>
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<tr>
<td></td>
<td>(0.009)</td>
<td>(0.020)</td>
<td>(0.011)</td>
<td>(0.019)</td>
<td>(0.015)</td>
<td>(0.019)</td>
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<td>Owns radio</td>
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<td>-0.011</td>
<td>0.010</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.011)</td>
<td>(0.020)</td>
<td>(0.016)</td>
<td>(0.022)</td>
<td>(0.017)</td>
<td>(0.020)</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Owns phone</td>
<td>0.002</td>
<td>-0.008</td>
<td>0.030***</td>
<td>0.022</td>
<td>0.025</td>
<td>0.059***</td>
<td></td>
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<td></td>
</tr>
<tr>
<td></td>
<td>(0.012)</td>
<td>(0.022)</td>
<td>(0.017)</td>
<td>(0.024)</td>
<td>(0.016)</td>
<td>(0.020)</td>
<td></td>
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<td>Food security</td>
<td>-0.001</td>
<td>-0.038**</td>
<td>-0.000</td>
<td>-0.027*</td>
<td>-0.014</td>
<td>0.024</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(0.007)</td>
<td>(0.016)</td>
<td>(0.008)</td>
<td>(0.015)</td>
<td>(0.013)</td>
<td>(0.017)</td>
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<tr>
<td>Weak</td>
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<td>-0.051</td>
<td>0.002</td>
<td>-0.041</td>
<td>0.080**</td>
<td>0.053</td>
<td></td>
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<tr>
<td>Dissemination</td>
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<td>(0.039)</td>
<td>(0.017)</td>
<td>(0.036)</td>
<td>(0.036)</td>
<td>(0.040)</td>
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<td></td>
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<td>NRM</td>
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<td>0.004</td>
<td>-0.003</td>
<td>0.003</td>
<td>0.043</td>
<td>0.033</td>
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<td></td>
</tr>
<tr>
<td>Councilor</td>
<td>(0.011)</td>
<td>(0.029)</td>
<td>(0.015)</td>
<td>(0.029)</td>
<td>(0.029)</td>
<td>(0.033)</td>
<td></td>
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<tr>
<td>Community meeting held</td>
<td>0.829***</td>
<td>0.623***</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<td></td>
</tr>
<tr>
<td></td>
<td>(0.011)</td>
<td>(0.016)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>SMS treatment</td>
<td></td>
<td></td>
<td>-0.083**</td>
<td>-0.123**</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td>(0.039)</td>
<td>(0.055)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Experienced problem</td>
<td></td>
<td></td>
<td>0.544***</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>(0.019)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Constant</td>
<td>0.407***</td>
<td>0.036</td>
<td>0.559***</td>
<td>0.311***</td>
<td>0.035</td>
<td>0.428***</td>
<td>0.440***</td>
<td>0.110**</td>
<td>0.419***</td>
</tr>
<tr>
<td></td>
<td>(0.017)</td>
<td>(0.029)</td>
<td>(0.052)</td>
<td>(0.016)</td>
<td>(0.038)</td>
<td>(0.055)</td>
<td>(0.015)</td>
<td>(0.048)</td>
<td>(0.054)</td>
</tr>
<tr>
<td>Observations</td>
<td>3405</td>
<td>3321</td>
<td>3321</td>
<td>3405</td>
<td>3321</td>
<td>3405</td>
<td>3321</td>
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</tr>
<tr>
<td>R²</td>
<td>0.001</td>
<td>0.718</td>
<td>0.037</td>
<td>0.002</td>
<td>0.487</td>
<td>0.056</td>
<td>0.005</td>
<td>0.248</td>
<td>0.018</td>
</tr>
</tbody>
</table>

Standard errors, clustered at sub-county level, in parentheses

*p < 0.10, **p < 0.05, ***p < 0.01
6.2 Propensity Score Matching

I estimate the average treatment effect of the treated (ATT) using three different variants of nearest neighbor (NN) matching: NN without replacement, NN with 5 nearest neighbors (NN5) and NN5 with a caliper of 0.003 (NN5 caliper).
Consistent with the DID estimations, I find no treatment effects for the general participation outcomes, which can be seen in the first three rows of Table 6. However, I find that the treatment effects on the local government participation outcomes contradict the findings in the DID method. While no significant treatment effects are found on the outcomes attended LG meeting and spoke at LG meeting, reported to LG is significant using all matching techniques. These results are consistent across the three matching methods used.

Using NN without replacement, all treated individuals are matched with one control. NN5 means exploiting the full sample since no observations are outside the common support and in NN5 with caliper, no matches are found for 14 treated observations.

<table>
<thead>
<tr>
<th>Dependent variable</th>
<th>Matching method</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>NN without replacement</td>
</tr>
<tr>
<td>Attended meeting</td>
<td>0.0063</td>
</tr>
<tr>
<td></td>
<td>(0.0299)</td>
</tr>
<tr>
<td>Spoke at meeting</td>
<td>0.0338</td>
</tr>
<tr>
<td></td>
<td>(0.0343)</td>
</tr>
<tr>
<td>Reported problem</td>
<td>-0.0529</td>
</tr>
<tr>
<td></td>
<td>(0.0353)</td>
</tr>
<tr>
<td>Attended LG meeting</td>
<td>-0.0170</td>
</tr>
<tr>
<td></td>
<td>(0.0223)</td>
</tr>
<tr>
<td>Spoke at LG meeting</td>
<td>-0.0021</td>
</tr>
<tr>
<td></td>
<td>(0.0300)</td>
</tr>
<tr>
<td>Reported to LG</td>
<td>-0.0846***</td>
</tr>
<tr>
<td></td>
<td>(0.0316)</td>
</tr>
</tbody>
</table>

Matching is done on the following variables: Age, Sex, Living standard, Education, Health, Social Capital, Owns radio, Owns phone, Food security, Weak dissemination and NRM councilor. SEs, bootstrapped using 50 replications, in parentheses. * p < 0.10, ** p < 0.05, *** p < 0.01

### 6.3 Heterogeneous treatment effects

It is possible that exposure to the scorecard information had heterogeneous impact on people with different characteristics and attitudes. In Hypothesis 2, I postulated that information will have a larger effect for individuals that: believe local government to be responsive to their voice; are dissatisfied with local government’s decisions and the quality of public services; have higher living standard; and are better educated. Table 7 reports heterogeneous treatment effects.
effects along these dimensions. The results are estimated with a triple difference approach where the parameter of interest is the DID estimator interacted with the potential source of heterogeneity in impact. I estimate the following model:

\[ Y_{ist} = B_0 + B_1(T_s \times POST_t \times H_{it}) + B_2(POST_t \times H_{it}) + B_3(T_s \times POST_t) + B_4(T_s \times H_{it}) B_5 T_s + B_6 POST_t + B_7 H_{it} + \gamma X + \epsilon_{ist} \]

In this model, \( H \) is a potential source of heterogeneous impact and the remaining variables are defined according to (1). \( B_2 \) to \( B_4 \) are interaction terms and \( B_5 \) to \( B_7 \) are linear terms. The description and coding of the variables in \( H \) are detailed in Appendix A2.

The first observed heterogeneous impact is found in individuals’ attitudes toward local government. Individuals who believe that local government is responsive to their opinions are more likely to attend and speak at meetings, but only at meetings called by local government. This heterogeneous treatment effect is significant for the outcomes spoke at LG meeting (p-value=0.026) and Attended LG meeting (p-value=0.020).

The second observed heterogeneous impact provides some support for the hypothesis that individuals that are unhappy with the current state of affairs are more likely to participate. Individuals who do not feel that local government decisions reflect their own priorities are more likely to participate in local government meetings, but not in meetings held by other actors. This is significant at the 10% level for the outcomes Attended LG meeting (p-value=0.098), Spoke at LG meeting (p-value=0.084) and Reported problem to LG (p-value=0.097). Also, Individuals that are less satisfied with public services are more inclined to report problems to local government. For Reported problem to LG, the effect is significant at the 10% level (p-value=0.099).

While no evidence of heterogeneous treatment effects is found for individuals’ level of education, a higher living standard appears to positively affect individuals’ likeliness to speak at local government meetings (p-value=0.073).
Table 7. Heterogeneous treatment effects of citizen attributes

<table>
<thead>
<tr>
<th></th>
<th>(1) Attended meeting</th>
<th>(2) Attended LG meeting</th>
<th>(3) Spoke at meeting</th>
<th>(4) Spoke at LG meeting</th>
<th>(5) Reported problem</th>
<th>(6) Reported to LG</th>
</tr>
</thead>
<tbody>
<tr>
<td>LG cares about me and my opinions</td>
<td>0.106</td>
<td>-0.211**</td>
<td>0.113</td>
<td>-0.151**</td>
<td>-0.010</td>
<td>0.070</td>
</tr>
<tr>
<td>LG decisions reflect my priorities</td>
<td>-0.028</td>
<td>-0.071*</td>
<td>-0.036</td>
<td>-0.061*</td>
<td>-0.044</td>
<td>-0.084*</td>
</tr>
<tr>
<td>Service satisfaction</td>
<td>0.058</td>
<td>-0.017</td>
<td>0.099</td>
<td>0.035</td>
<td>0.145</td>
<td>0.200*</td>
</tr>
<tr>
<td>Living standard</td>
<td>-0.067</td>
<td>-0.067</td>
<td>-0.105</td>
<td>-0.065*</td>
<td>-0.001</td>
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<td>0.006</td>
<td>0.008</td>
<td>-0.002</td>
<td>-0.012</td>
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</tbody>
</table>

Reported coefficients are the DID estimator interacted with the term at the left. All specifications include the following control variables: Age, Sex, Living standard, Education, Health, Social Capital, Owns radio, Owns phone, Food security, Weak dissemination and NRM councilor. Design weights are applied. Standard errors, clustered at sub-county level, in parentheses. *p < 0.10, **p < 0.05, ***p < 0.01

7. Conclusion

Providing citizens with enough information to assess the performance of their elected politicians is widely seen as prerequisite for accountability (Keefer and Khemani, 2005). This thesis does not question this assumption but argues that information alone is not sufficient to strengthen the accountability relationship between citizens and their elected politicians.

In this thesis, I test the effects of providing citizens with information about the performance of their locally elected politicians and ask whether such information induces more citizens to actively participate and demand better public services. Contrary to my main hypothesis that the scorecard dissemination will have positive effects on participation, I find that information is not only insufficient to activate citizens but leads to disengagement in the participation forms of attending community meetings and speaking at community meetings. This negative effect of the scorecard dissemination is, however, only found on meetings called by local governments. I argued that the treatment effects should be larger on the local government participation outcomes due to their more straightforward connection to the ID treatment, which is also found to be the case. The results using PSM, however, cast some doubts on these findings by showing that the treatment effect is negative and significant on citizens’ reporting of service delivery problems to local government while no treatment effects are
found on citizens’ participation (attending and speaking) at local government meetings. I argue that DID provides a better estimation strategy in this setting where data limitations are a factor. This is because DID unlike PSM allows for some selection on unobservables.

A potential explanation for the negative findings is that the scorecard scores fell short of citizens’ expectations of how well their district councilors perform (Gottlieb, 2016). If citizens overestimates the performance of their district councilors, the expected benefits of participating might be reduced, leading more citizens to not participate (Banerjee et al., 2011). Low scores might also make salient features of district councilors that exclude citizens from the decision-making process, which could lead more citizens to disengage (Buntaine et al., 2018). For instance, if district councilors score low on legislative duties, they are unlikely to provide an efficient channel through which citizens can influence service delivery outcomes. That is, if citizens perceive a councilor as unlikely to influence the work and decisions of the district council, citizens might refrain from seeking to influence this councilor. If being exposed to low scores makes more citizens reluctant to participate, this effect should be stronger for attending meetings and speaking at meetings since these outcomes are likely to require more effort than reporting a service delivery problem. This is also found to be the case. Unfortunately, data limitations prevent me from further investigating low scores as a potential explanation of the negative effects.

An important driver of the negative results appears to be the relatively widespread perception among citizens that local governments do not care about them or their opinions; at baseline (2013), 60% of respondents had this perception. Investigating heterogeneity in treatment impact among citizens with different attributes, I find that individuals who do not believe that the government cares about them or their opinions are less likely to participate. In contrast, I find that individuals who do not feel that local government decisions reflect their priorities are more likely to participate. These seemingly contradictory findings suggest that attempts to influence local government decisions are more likely for citizens that perceive local government as an actor that is benevolent and care about the citizens but also as an actor whose priorities and decisions could be improved.

These results point to the need of raising citizens’ esteem for their own ability to influence local government decisions. This could be achieved, for instance, by education or by focusing on facilitating collective actions for citizens that are more unlikely to participate (Buntaine et al., 2018). While increasing government responsiveness is goal of many transparency
initiatives (Gaventa and McGee, 2013), the results suggest that this could trigger a virtuous circle where government responsiveness and citizen participation reinforce one another. Furthermore, the results highlight the need to further raise citizens’ expectations of what is an acceptable performance of politicians, which could contribute to more citizens participating.

The most important limitation of this thesis is that I am unable to assess to what extent individuals in the treated areas actually were treated (received the scorecard information). While on average 40 people attended the two information dissemination meetings at parish level in 2013 and 2014, these individuals should not be able to reach all citizens in a parish even if they have prominent positions in society, such as being religious or civil society leaders. Instead, the experiment relies on a sufficient buzz around scorecards having been created throughout the communities. The information dissemination tools (flyers, posters, calendars) need to have reached a sufficient number of people and, importantly, citizens need to have talked about and discussed the scorecards with each other. That ACODE had been working in the districts subject to the experiment for at least two years prior to the treatment should have made some people aware of who stands behind the information and hence give some credibility to it. Chances of a good outreach are also improved by the fact that there were two dissemination rounds. If an individual received a flyer about the scorecards the first time and paid no attention to it, she might develop some curiosity the next time she hears about it.
References


Appendix A. Variable descriptions

A1. Control variables

*Sex* is coded as 0=female, 1=male.

*Age* is the age of respondent, measured in years.

*Education* is the highest level of education completed by the respondent on a 16-point scale.

*Owns phone* is a binary variable taking the value 1 if the household of the respondent owns a phone.

*Owns radio* is a binary variable taking the value 1 if the household of the respondent owns a radio.

*Food security* is calculated as the average of four questions: In the past 7 days how often have you had to: Rely on less preferred and less expensive food? Borrow food, or rely on help from a friend or relative? Limit portion size at mealtimes? Restrict consumption by adults in order for small children to eat? Reduce number of meals eaten in a day? The respondents were given 5 identical choices to each question, coded as follows: 0=never, 1=rarely, 2=sometimes (three to ten times), 3=often (more than 10 times), 4=Always (every day).

*Social Capital* is dummy variable taking the value 1 if the respondent answered yes to the question: If you suddenly needed 20,000 UGX to pay for a health treatment, would you be able to borrow money from family/ friends?

*Living situation* captures respondents’ perceptions of how well-off they are, coded on a 4-point scale. The question asked: “Which of the following statements best describes your household’s situation over the past 12 months?” 1= “Doing well: able to meet household needs by our own efforts and making some extra for savings or investments”. 2= “Doing just okay/ breaking even: able to meet household needs but with nothing extra to save or invest”. 3= “Managing to meet household needs, but we have to sell productive assets or sometimes rely on help from others”. 4= “We are unable to meet household needs by our own efforts. We depend on support from community or government”.

*Health problems* is dummy variable taking the value 1 if anyone in the household of the respondent has experienced long term health problems in the last three years.
NRM councilor is a dummy variable taking the value 1 if an individual lives in a sub-county that is represented by a regular councilor from the National Resistance Movement at the district council.

A2. Variables used for heterogeneous treatment effect assessment

Education and living situation follow the description and coding above.

Local government cares about me and my opinions is a binary variable taking the value 1 if the respondent answered yes to the question: “Do you agree with the following statement: The local government cares about me and my opinions?”

LG decisions reflect my priorities is based on the question: “To what extent do you feel that the decisions of the Government in the District and sub-country reflect your own priorities?” Respondents were given the options: 1=“Never”, 2=“Almost never”, 3=“Only in some areas”, 4=“Very much, to a large extent”, 5=“Absolutely, always”.

Service satisfaction only concern health and education services and is based on the questions: Overall, how satisfied are you with the education services you use? Overall, how satisfied are you with the quality of service you received in your most recent visit to the health centre or clinic? In these two questions the respondents were given identical choices: 1=Satisfied, 2=Fairly satisfied, 3=Dissatisfied.
Appendix B. Tests for propensity score matching

Figure B1 Propensity scores before and after matching using NN5
Figure B2. Propensity scores before and after matching using NN without replacement

Figure B3. Propensity scores before and after matching using NN with a caliper of 0.003
The table compares means of the untreated and treated group for each variable included in the PSM models. This is done by t-tests for equality of means. For NN5, the means of the treated and untreated group are reported while only the t-test is reported for NN without replacement (NN no replace) and NN with a caliper of 0.003 (NN caliper). Significance level of t-test indicated by * < 0.10, ** < 0.05, *** < 0.01.

<table>
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<th>Variable</th>
<th>Unmatched Mean</th>
<th>Matched Mean</th>
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<th>T-test Matched</th>
<th>T-test NN no replace</th>
<th>T-test NN caliper</th>
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Table B1. T-tests of differences in means for the different matching techniques