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**Computational Representation of Social Complexity For Decision Making:**

**The Case of Climate Change, Migration, and Social Conflict**

***Abstract:***

*This chapter illustrates how computer simulations can help us foresee critical situations in populations where climate change may cause water scarcity. In particular, we exemplify a city in the central Andes of Peru whose conditions related to deglaciation, urbanization, and population growth, may bring water scarcity. The main technique used is agent-based modeling, which will combine quantitative and qualitative information. The results show possible scenarios of water scarcity leading to drastic migration and social conflict, a situation not yet evident as a public problem. This work provides recommendations for this likely and undesired scenario.*

*Keywords: Social complexity, agent-based modeling, climate change, migration, conflict.*

**1. Introduction** *(should cover the rationale for study or practice, the research aim and/or research question(s), research or practice gaps)*

Social problems are complex. That is, combine environmental, technological, psychological, social, political, and cultural issues. This complicates policy and political decision making as one branch alone of the government cannot solve public problems effectively. However, involving several branches and multilevel decision makers will not make the discussion more efficient.

This chapter presents social simulations via agent-based modeling (ABM). As Macy and Willer (2002) propose, ABMs will allow the social scientists to change their modeling efforts from representing societies as interactions among variables into a different paradigm where social life can finally be represented as interactions among adaptive agents who influence one another in response to the influence they receive.

This chapter exemplifies the use of ABMS to understand the interaction of climate change and population growth in a small-sized city. The ABM will allow us to envision possible futures as climate change trends affect water availability, the growth of the population urbanizes neighboring rural areas, and people need to decide in the future whether to stay or leave.

Water scarcity and droughts have a slow onset (Singh et al., 2021), so as time passes by, most urban and rural settlers do not pay much attention to this issue. However, if we look into the future the absence of water may affect the decision of settlers in the urban and rural areas in different ways that need a preliminary answer:

* Is it possible that drastic migration patterns appear as people do not find water?
* Is it possible the emergence of social conflict in the area among the people that can not migrate and find themselves in a place with no water?

The purpose of an ABM is to serve as an input for anticipatory and participatory political decisions (D’Aquino et al., 2002). The presence of different actors with different goals and beliefs, and under stress due to the environmental conditions makes it challenging for mainstream modeling techniques based on variables to provide the right guidance, so this example hopes to contribute methodologically to the coming discussion of the future of the area selected. There are certainly many computational modeling approaches—e.g. system dynamics or discrete systems simulations—that deal with organizational complexity. However, key concepts such as learning and emergence can only be modeled, arguably, through the use of ABMs (C. A. Cioffi-Revilla, 2014; Epstein & 2050 Project, 1996; Gilbert & Troitzsch, 2005; Miller & Page, 2007). As with all modeling techniques, ABMs need to capture the most basic variables and processes that could explain a particular phenomenon, while making all building assumptions explicit and transparent, and producing outcomes valid, and of interest to the scientific and policy community. In contrast to other techniques, ABMs are virtual laboratories that allow their components to have different behaviors, reactions, and to be aware of only limited information. In particular, ABMs are useful in the social sciences when a particular social theory can be enacted through coding. You all can then see how the model behaves as the building parameters of the theory are manipulated (Epstein & 2050 Project, 1996; Miller & Page, 2007). Besides, ABM allows for the inclusion of complementary theories and models to carefully enrich the original theory, allowing the creation of different scenarios.

Climate change is a major social concern that needs profound reflection for the future of humanity; particularly if it affects the availability of water (Robinson & Klobucista, 2023). But even if climate change were not a concern, urban population growth by itself will challenge the best efforts in water management (Alfarra & Turton, 2018).

The application presented is based on the area that depends on the water from the Shullcas River in the central Andes of Peru. The Shullcas river is a proglacial river that originated from the melting of the Huaytapallana tropical glacier, a kind of glacier more likely to disappear if global warming keeps its current trend (Willige, 2023). The area includes rural and urban settlers that make exclusive use of the water from that river. The area has a positive population growth, with a marked tendency to urbanize the rural area located on the Huaytapallana ’s foot and skirt. Given this context, this example serves the current studies on computational representation of social complexity:

* First, the object of study is not one particular kind of population (as it is the case in most studies) but it studies simultaneously rural, and urban and peri urban populations; so the water issues will be of concern to different kinds of groups with different concerns on water, which may be the starting point for inter group conflict (Bar-Tal, 2011).
* Second, this ABM will include the effect of a glacier in the water supply for the social system, facilitating the analysis of similar systems by separating the two main kinds of precipitation: rainfall and snow melt.
* Third, the watershed under analysis may be a representative case for the whole Andes chain, as it is at the center, among the dry biggest cities on the west and the sparsely inhabited Amazon jungle to the east.
* Fourth, it integrates and harmonizes current modeling efforts on the glacier melting trend, the hydrology dynamics and the social behavior; making our work more challenging as these systems have never been integrated before.
* Finally, and related to the previous point, the issue is not of particular concern neither to most of the population nor to the political authorities; it looks as though nobody wants to touch the situation as it may collide with the current way people live and benefit. This work also continued with the best practices in computational social science (CSS) research used by the previous studies, as it integrated the complex adaptive systems approach(Miller & Page, 2007), the CSS consolidated techniques (C. Cioffi-Revilla, 2017), and ethnographic work (Crate & Nuttall, 2009).

**2. Literature review (background) or problem background/ context** *(pertinent literature is reviewed, up-to-date sources used, research or other best practices reviewed and critiqued, important terms defined)*

Research interests dealing with modeling climate change are varied, but they tend to have a common interest in rural populations, especially vulnerable and located in developing countries. Ziervogel et al. (2005) focuses on modeling decision-making and forecasting of farmers in Lesotho, in an scenario where official forecasts, besides poorly done, are not diligently nor appropriately disseminated towards these end users, making them more vulnerable as loses due to poor information endanger the farmers’ survival. Similar research is found in Bharwani et al. (2005) for smallholder farmers in a village in Vhembe district, Limpopo Province, South Africa.

Modeling using ABMs in more comprehensive approaches are found in Kniveton et al. (2011) and Hailegiorgis (Hailegiorgis, 2013). The former has Burkina Fasso as its case study, investigating the role of the environment in the decision to migrate using scenarios of future demographic, economic, social, political, and climate change in a dry land context. In that work, it is found that change to a drier environment produces the largest total and international migration fluxes when combined with changes to inclusive and connected social and political governance, while the lowest international migration flows are produced under a wetter climate with exclusive and diverse governance scenarios. And the latter focuses on the South Omo zone in southern Ethiopia, modeling how the current surge in large-scale land acquisition might affect rural livelihood. In this work, Hailegiorgis found that (i) the occurrence of drought affected their adaptive capacity and forced migration; (ii) increasing the magnitude of expansion of large-scale land acquisition aggravated dispossession and increased migration of rural households; and (iii) capacity building and relief support in the time of extreme events minimized migration.

As it is clear so far, climate change stresses the place where human activities are carried out, thus deteriorating the quality of the environment of these populations forcing them to make different decisions, which, as Hirschmann proposes (1970), can be categorized as exit, voice and loyalty. The previous cases speak of exit, as people emigrate looking for better conditions as the climate threatens their well-being; but the confrontational and loyalty dimensions in social complexity issues also need to be analyzed under climate change effects. However, loyalty and voice seem to be less treated from the global computational social science community, which should be considered quite shocking as reports related to climate change and conflicts are abundant (Hendrix & Salehyan, 2012; Hsiang et al., 2013; Steinbruner et al., 2013). Related to conflict and climate change, Piontek (2010) studies the Nile River, which serves 10 countries in Africa, seeking to understand, via computational modeling, how the interaction between humans and the environment may lead to conflict or cooperation. In this work, the author found that climate change can alter conditions for conflict and cooperation, but it might not be the immediate cause for those issues. Nevertheless, the most outstanding work related to climate change and the emergence of conflicts can be found in the works produced at the Center for Social Complexity at George Mason University funded by the Multidisciplinary University Research Initiative (MURI) and Cyber-Enabled Discovery and Innovation (CDI) programs (C. Cioffi-Revilla, 2010; Hailegiorgis et al., 2010; Kennedy et al., 2010). They focus on the complex interaction of pastoral groups with their environment and other emerging external actors in east Africa finding that increased seasonal rainfall variability and droughts create tremendous stress on pastoral groups and challenges their long-term resilience and adaptive response mechanisms, thus concluding the population’s relation to the carrying capacity seems to be the major factor affecting cooperation and conflict.

**3. Methodology: tool(s) or method(s) presentation for problem-solving or decision-making**

The ABM presented is very simple. It is built to capture the relationships that answer our questions and to be useful for policy discussion. Thus, we have a basic model composed of two systems, the social and the natural one. Both systems interact and the ABM will see if migration and conflict become more likely as time goes by. This is represented in Figure 1.

<FIGURE 1 HERE>

As it can be seen in Figure 1, the natural system focuses on the water supply, as it is the key trigger factor for the social issues of interest in this chapter. The current studies of the area propose both a decreasing supply of rainfall, and forecast the complete retreat of the glacier, which will affect the balance negatively (Carlos & Grijalva, 2012; López-Moreno et al., 2014). One obstacle for a more precise model is the lack of information on the snow melt supply; however, this value was bounded based on expert opinions**1**. On the other hand, water demand comes from the social system, which is represented by two subsystems, rural and urban. The basic process in both subsystems is population growth**2**, that is, eventually there will be not enough water supply for the population’s demand. We explore emigration as a particular response (other responses have not been considered at all): most Andean cities have negative population growth, but the area under study has a positive one - influenced by immigration, so the positive case makes it more politically relevant. Other demographic information is fed into the model (working, educational and marital status) to model the agent behavior.

Hypothesis Operationalization

As this work proposes that the research questions have affirmative answers, the following operationalization of the hypothesis as been considered, see Figure **2**:

<FIGURE 2 HERE>

The variables present a simple yet comprehensive process to study the potential for conflict and migration:

* The potential outcomes are chosen as the representation of our framework exit, voice and loyalty (Hirschman, 1970). In this case, exit is represented by drastic migration; loyalty by the permanence in the area (not shown); and voice by expected reaction from the frustration felt by urban people, or the relative deprivation felt by rural people.
* The drivers clearly represent the current trends in the area, both in the social and natural dimensions. These trends are not altered in the ABM. The assumption that the trends can be extended into the future are based on the historical population growth and immigration. The water balance trend represents the other important driver, which will be represented by water supply and demand. Water supply will be based on the contribution of the glacier Huaytapallana.
* The direct consequences are the clear results of the unaltered trends. The ABM includes the two most clear direct consequences supported by the literature and field work (Haller, 2012; Haller & Borsdorf, 2013; Ho & Milan, 2012). Water scarcity is a first expected outcome as the water balance and population growth trends remain unaltered. Water scarcity will become a major problem, as explained in (Carlos & Grijalva, 2012), which combined with population growth and, peri urbanization may exacerbate emigration and conflict (Haller, 2012; Haller & Borsdorf, 2013; Ho & Milan, 2012).
* The conditional factors are processes that work at the individual level and are the main sources of heterogeneity in the model. Every individual will need to find a particular response based on these factors. The factors have been selected from the classical work of Maslow & Frager (1987), and its application in agent-based modeling to understand conflictive situations (Watkins et al., 2008). According to these individual level factors, the ABM has implemented several key processes: (i) a Bayesian mechanism to compute and update expectations, (ii) information about local resilience to water scarcity based on field work in the area; (iii) the assigning of capacities to agents (citizens) based on census data. These competences reflect closely what determines the potential outcomes (Watkins et al., 2008) based what the needs an agent needs to cover: water (physiological needs), work (safety needs), family (belonging needs) and education (for esteem and self-realization needs).

Field work

We decided to organize an ad-hoc collection of information in situ as the ABM required some information for the agents not available neither on the literature nor the data. We needed to have a clearer idea on how the agents will react as water will become scarce in the rural and urban area. For that reason, we organized two research expeditions: One in the urban area and another in the rural area; Figure 3 shows the area of study and the area covered in each expedition.

<FIGURE 3 HERE>

The rural area has the longest extension of land but is also much less populated than the urban (around 10% of the urban population). The team made unstructured interviews to find out about: (i) the awareness of the water scarcity in the future; (ii) their need to migrate in case of water scarcity; (iii) the possibility of taking some political action. The main findings were (i) person after person affirmed that drastic migration is not an option for the rural; (ii) the increasing presence of immigrants urbanizing the rural area represented an unappealing fact. The study also confirmed people were very aware of the progressive lack of water, but they believe this issue will be of more concern in the urban area. A couple of additional important facts were unveiled. First, they are sure that the urban needs of water will collide with the rural needs. They are not sure how much water the growing city may need in the future. Also, they are confident local authorities will not consider their opinion if the city needs more water. Second, they believe this generation of rural people will do their best effort to make a living in the rural area; but they are not very sure what would be the reaction of the future generations. They consider that the jungle represents an attractive destination for young unemployed people but are afraid the attraction comes from easy-money and illegal activities.

The peri urban expedition visited 65 families living in the limit between the city and the rural area. A survey and interviews were designed and applied. The team designed a polyetapic sampling design to include a representative sample of households of migrants and non-migrants in the urban settlements located along the peri urban area. This expedition wanted to find the same information as the previous one. The information collected in the peri urban area informed: (i) urban people have migration among their alternatives in case of lack of water; (ii) the urban settlers will look for a place nearby to move in (they may need to get land closer to the rural area). It is worth noticing that all who said they would never migrate are immigrants; (iii) all believe that they will feel frustrated if they must stay in place with water scarcity. This expedition found that people in the peri urban area did not start living in that zone but renting a place downtown, which was expensive, smaller and overcrowded, and, progressively, they moved (renting or buying) into the peri urban area as it represented a cheaper place to live. The strategy they followed to find a better place was a system of references, where family and friends were sharing their experience living outside the city center, even with lack of public utilities in the beginning. The algorithm will use this strategy to search for a location to rent, asking social networks of families where the water is still available.

Modeling building

The ABM simulates 50 years. Every time a complete cycle of the simulation is run, the history advances six months. So, this model will only run for 100 cycles. As each cycle represents six months, a cycle represents a season, which is either the dry season (that goes from May to September) or the rainy season (that goes from October to April). Figure 4 reflects the flow of the code each cycle or season.

<FIGURE 4 HERE>

Figure **5** is another flowchart representing the decision making of the agents. For the rural agent, as it is shown in that figure, the decision flow starts when the agent detects scarcity. That detection is a particular computation following a Bayesian approach, which was assumed as the mechanism to update beliefs. Once the agent detects scarcity, he will look for a place to move. As long as the agent detects scarcity where the agent is staying, the agent will move. The moving will end when the agent reaches the resilience limit. If the rural agent can migrate, the rural agent will migrate to the urban area. Agents that could not migrate are candidates to feel relatively deprived.

<FIGURE 5 HERE>

Figure **5** also shows the flowchart representing the decision making of an urban agent. Similar to the rural one, the decision flow starts when the agent detects scarcity following a Bayesian approach. Once the agent detects scarcity, he will look for a place to move. If the agent detects scarcity where the agent is staying, the agent will move. The moving will end when the agent reaches the resilience limit. If the urban agent can migrate, the urban agent will migrate either to the coast or the jungle, depending on the agent’s characteristics (employed, educated, marital status); the more vulnerable agents will go to the jungle. Agents that could not migrate are candidates to feel frustrated. Once the agent is frustrated, the agent will try to connect to other agents in the same situation. These connections make a social network of frustrated or angry agents. A link is destroyed when a member of the network of frustrated agents migrates.

**4. Results: problem-solving or decision-making based on selected tools or methods/ or best practice(s) presented**

Rose (1993) believes anticipation allows policymakers to forgo the necessary rigors of empirical evidence, as anticipation is not a scientific endeavor but in fact a political tool that can be used when facing uncertainty and novelty. This ABM results may help anticipate a critical situation:

* Take one. The model has been verified and validated. All the population is situated in the year zero and the model is ready to run (after last census in 2017). The population count reflects a proportion of actual population. The areas are also proportional. The Glacier Huaytapallana area in in the year zero is represented by the zone in white, the rural settlers inhabit the green area, and the urban agents are in the gray area. The water balance is not an issue yet, the water supply is enough for the population water needs. See Figure 6 (a).
* Take two. As time passes by, the glacier keeps melting following the identified trend. The pink patches near the white patches represent the zone retreated. Even though there are no water issues in this moment, people started slowly moving away from the more densely populated area (left gray area) into the peri urban area. The main driving force for this is the population growth. See Figure 6 (b).
* Take three. The glacier has retreated almost completely, and the water balance for the urban area is now negative. The pink patches in the urban area are affected by the drought during the dry season. The drought is felt progressively from the highest risk zones (leftmost gray zone). If an agent is living in a patch that suffers drought, he may need to move away, depending on his beliefs of future scarcity. There are some urban agents that could not migrate when their resilience was reached. They are already organizing into a network. The peri urbanization process continues but no rural is feeling relatively deprived yet. See Figure 6 (c).
* Take four. The Huaytapallana has melted completely. This has made the situation even worse during the dry season. Urbans are populating more the peri urban area, and the increasing presence of immigrants has started making rural people feel relatively deprived. The rural agents with a bigger size represent those rural agents. Even though the migration into the coast or the jungle has been massive, there are still many people feeling frustrated living in the area. Not every frustrated urbanite is connected to one another, but there are many network components and cliques everywhere in the city. The simulation is about to finish after representing 50 years. The potential for conflict and migration turn into a fact in the model. See Figure 6 (d).

<FIGURE 6 HERE>

Let’s try some answers to our initial questions.

Is drastic migration possible?

Related to migration, Table 1 informs that people are moving away. However, one can see that urban people did migrate, while rural people did not, even though the possibility to migrate was open to both groups. Migration started at period 54 (year 27), the jungle was by far the most preferred destination. Migration was not immediate for the simulation, it started after 27 years (54 periods), and five years later the most dramatic migration happened. If the demographic conditions held true, the migration to the jungle would be the most alarming.

<TABLE 1 HERE>

Is social conflict possible?

Table 2 offers several measures of the network of angry people in the area, which was used to infer the possibility of conflict. That table should be read in tandem with the Figure 7. The density is very low, but it is due to the high number of components, however, the clustering coefficient is not very low, because the giant component is big. Altogether, you can infer that angry people are grouping. Then, this situation represents a likely situation for conflicts to emerge (Cárdenas, 2013; Levina & Hillmann, 2013; Smith & Vivekananda, 2009). Form Table 2 we also know that the simulation ended with an immigrant population representing 25% of the total urban population; that is, after 100 periods (50 years) (the immigrant population at the beginning of the simulation was 5% of the native population in the urban area based on official data(Instituto Nacional de Estadística e Informatica & Organización Internacional para las Migraciones, 2015) a fact that adds to the explanation of possible conflict, as less ‘familiar’ people are found in the neighborhood. It is also clear the share of rural people angry is far less than the share of urban people angry; that would mean that the urban area has more people inclined to satisfy their needs for water (housing) without moving away.

<TABLE 2 HERE>

<FIGURE 7 HERE>

**5. Conclusion:**

* 1. *Implications for Practice*

This model, despite the real data that has been used to feed it, is still more a political tool than an engineering one. Its purpose is to show a potential scenario instead of a clear forecast, but still gives a clear picture of what might happen if no decision is made to secure water. This model’s goal is to raise enough awareness that triggers real political action. With that in mind, there are some recommendations on the next steps policy makers should follow:

Make sure people in rural area have enough rights and mechanisms to stay in their lands in good conditions. The tendency of immigration may not diminish in the short run and rural people have no intention to leave the area. To avoid relative deprivation emerge, policy makers from the central government need to assist rural populations to adapt their farming practices; regional planners need to determine buffers that separate urban from rural areas as well as assure better educational conditions, while local governments need to stop giving building permits in the buffer area.

Create mechanisms to ensure political participation of rural people. Rural people are politically underrepresented. They only have local committees that regulate their internal activities in the community; but they have no real presence in the regional decision making. Local and regional governments should make sure that rural people have a voice to avoid future discrepancies. Policy makers must keep in mind that the model is constantly recommending increasing the water reserve from the rainy season, which will need investment and the use of arable and/or herding land. In any case, rural communities should not feel they are affected to favor the urban.

Improve participatory water governance. The distribution of water between rural and urban areas is currently a process that represents a constant debate. From the interviews, rural people believe that the city is taking more water than they should from the Shullcas during the dry season and that the rural needs are relegated due to the pressure of the urban majority. Water management has many actors involved, including two Ministries from the central government (Agriculture and Environment), but the technical criteria adopted so far is biased toward the needs of the urban area.

Do not hide the crisis but show a plan. The dramatic situation must be shared with the population, but in a message that shows the local governments have a plan that needs urban people to alter their inefficient use of water, or to learn routines that avoid unnecessary use of water. If the water demand decreases in the long run, make sure this process gets as slow as possible. For that, different campaigns should be promoted at elementary and high school level to be more efficient using water in the future generations; organize contests at college level to promote local inventions; and organize neighborhood contests to demonstrate how to achieve efficient practices in the use of water.

Seek collaboration from civil society, especially research institutions. The highlands of Peru have always been arenas for NGOs to operate. Ideally, NGOs can detect social problems and work with the community to empower their actors; but this time the challenge is different, as it deals with the resilience and adaptive capacity of people facing high uncertainty of the future environmental conditions, and the reactions their neighbors themselves will have in those critical moments. In this situation, the production of more knowledge is needed as well as interdisciplinary debate. The government has serious limitations to hire people permanently, but the local public University should not remain as the forgotten partner that could make all the difference. The local University has enough resources to fund important programs that can keep updating the social and natural situation for more informed policy making and modeling, and which should create mechanisms for a two-way knowledge transfer, that is to share scientific knowledge and collect traditional knowledge.

Be careful of easy solutions. This work has suggested many times the saving of water. This is a decision that needs to be very well planned. To start, a huge reservoir for the area seems to be a good solution, but decision makers must be aware of the Huaytapallana failure (Dorbath et al., 1990). This failure turns the zone very sensible to earthquakes, and a huge reservoir represents a great potential for disaster. The saving of water will need a lot of work from the political class and the people themselves. From the top down, the infrastructure to save or become more efficient should be built; and the urban and rural settlers have the responsibility to be more efficient using water. Another important way to secure water will be a better management of the groundwater, which currently is used without detailed knowledge of its quantity nor quality, and without any recharging policy.

* 1. *Limitations*

Every model needs assumptions and facts, those facts can be qualitative or quantitative, and this agent-based model has dealt very well with facts and assumptions. However, the most urgent areas of improvement can be:

Have a better sub model of what an agent could do when it predicts water scarcity is important. In our model, the agent looks for another source of water, and when the agent finds a better place, it abandons its home and moves to another place within the system and keeps doing that until his desire to stay in the system is surpassed. However, we would have preferred to have information on neighborhoods and its economic capacity, to implement an algorithm considering that information, but that data does not exist. An important investment would be needed to produce that data.

This model assumes that the coast or jungle are the main destinations for the emigrants, but coast in fact represents “big cities in the coast” and the jungle represents “good places to go if you are a risk taker and you are not planning to bring your family”; although the answers from the interviews mentioned these two places it would be important to expand the model and see how the new arrivals are received and what new issues arise in those destinations. The answers also mentioned “foreign countries”, but that destination has not been included (but “coast” could represent it).

It would be important to consider more strategies of the agent according to its economic capacity. The economic capacity is unknown at agent level, so further work on this should be done. It is also important to conduct a more detailed analysis on water regulation from the government to get the agents to reduce consumption; since Peru is a country where no political authority wants to alter the price of water (water is very cheap in Peru), a different regulation mechanism should be thought of following a participatory approach.

A different programming platform could be important to consider. At this point, we have reached the capacity of the program NetLogo (Wilensky, 1999; Wilensky & Rand, 2015) but if this model becomes more computing-intensive, there is a clear need to migrate the model into more sophisticated platforms such as Mason (Luke, 2011; Luke et al., 2005, 2015), Repast (North et al., 2013), or Gama (Taillandier et al., 2019). In fact, the code is ready to be sent to real maps, and see the simulation in a more realistic setting, but those other ABM platforms are needed to make the conversion useful.

* 1. *Future Directions and Perspectives*

The natural barrier to anticipation is a reactive mindset, which is very common in policy making (Torjman, 2005). For sure, one cannot blame decision makers for being reactive as generally the political institutions are a set of rules that have been conceived based on what is known. There may be many examples of catastrophes that could have been avoided if the right anticipatory measure had been taken previously, but it is also true that the legal systems generally lack means to make decision makers liable for their lack of planning; and only electoral means are a way to express the general discontent for their lack of proactiveness. However, electing a new or better political leader does not undo the damage and suffering of the affected people. A reactive system of policies is not completely bad and could even be considered an efficient way of spending public money. However, when combined with other factors, the whole situation can become catastrophic.

A first direction to consider should be understanding the difficulty of decision making in extremely centralized systems or weakly decentralized. A centralized system will react only after every institution beneath has considered that there is a need for anticipation, and a weekly centralized system gives the local institutions the illusion of decision making, when there is a regulation that will force the local authority to deal with a complicated set of institutions. The Shullcas and the Huaytapallana may be clearly affected by this situation.

Another negative ingredient to reactive policymaking is symbolic and superficial proactiveness. Particularly when dealing with this climate change issue, programs and organizations are created, technical documents are produced, presentations and dinners are offered, but at the end of the day there are no real measures beyond declarations of concern.

The level of economic development is also a key factor in understanding the ability to cope with water shortages, as more developed countries have more technological and financial resources to deal with drastic environmental issues. What can poorer countries, or localities do? they have meager funds and often suffer through political instability that constrains implementation of effective and long-lasting policies. However, economic development has not stopped advanced economies like the USA to be reactive in many cases in many different issues like the Challenger accident, the 9/11, Katrina hurricane, and so on (for more stories see Bazerman (2004)). Then, poor leadership affects a reactive system of policies. A poor leader plays safe, and tends to leave difficult situations to experts, framing complex issues as technical. As explained by Heifetz and Linsky (2002), once leaders frame a complex problem as technical instead of complex-adaptive, the political system just does “routine management” instead of “change management”. This explains that the reports on the watershed appeared around 2010, and all of them were very technical; and since then, no more knowledge on this area has been produced from the government.

1. **Acknowledgment (s) (if any)**
2. **References (APA Style 7th ed.)** *see the attached template*

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1. **Biographical notes of Authors (100-150 words)**

Jose Manuel Magallanes is a full Professor in the department of social sciences and director of the Institute for Social Analytics and Strategic Intelligence at Pontificia Universidad Catolica del Peru (PUCP). He has had part-time appointments at Universidad Nacional Mayor de San Marcos (UNMSM), University of Washington (Seattle), and University of Massachusetts (Amherst). Professor Magallanes holds a BSc in Computer Science (UNMSM), a M.A. in Political Science (PUCP), a PhD in Psychology (UNMSM) , and a PhD in Computational Social Science from George Mason University.

**Tables and Figures:**

*Table 1.*Migration-related results

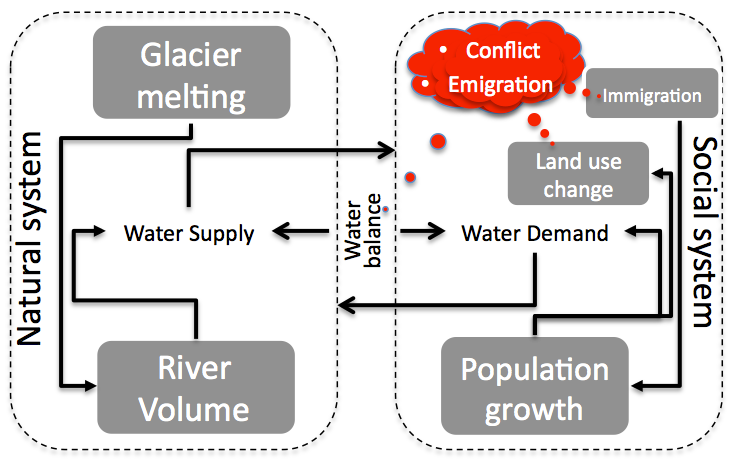
|  |  |
| --- | --- |
| *Variable* | *Value* |
| Urban people that moved to the Jungle | 2662 |
| Urban people that moved to Lima | 356 |
| Rural people that moved | 0 |
| Moment Migration started | 54 |
| Max amount that emigrated in a year to the Jungle | 337 |
| Max amount that emigrated in a year to Lima | 54 |
| Max amount that emigrated in a year | 391 |
| Moment when Max amount that emigrated in a year was reached | 64 |
| Moment when Max amount that emigrated in a year to the Jungle was reached | 64 |
| Moment when Max amount that emigrated in a year to Lima was reached | 64 |

Source: Author.

*Table 2.* Conflict-related results

|  |  |
| --- | --- |
| *Variable* | *Value* |
| Max amount of angry urban people reached | 2497.00 |
| Density of Angry Network when max amount of angry people reached | 0.01 |
| Average clustering coefficient of Angry Network when max amount of angry urban people reached | 0.41 |
| Number of network components when Max amount of angry urban people reached | 491.00 |
| Moment when Max amount of angry urban people reached | 96.00 |
| Share of immigrants in urban area | 0.25 |
| Rural population when max amount of angry urban people reached | 569.00 |
| Urban population when max amount of angry urban people reached | 3116.00 |
| Moment that first rural started feeling relatively deprived | 80.00 |

Source: Author.

****

*Figure 1.* Basic model abstraction. The red elements represent the potential outcomes if the other elements keep the current trends

Source: Author.

A diagram of different factors

Description automatically generated

*Figure 2.* Operational Variables in Hypothesis.

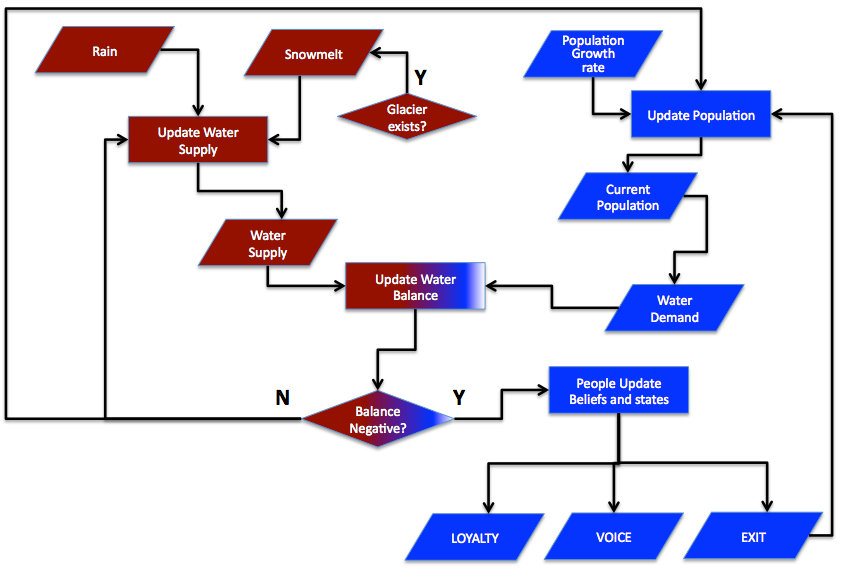
Source: Author.

A map of a river

Description automatically generated

*Figure 3.* Map of Field work. The black oval is the area covered by the urban expedition and the red oval the area covered by the rural expedition.

Source: Author.

**

*Figure 4.* Process Overview. Color differentiates type of system (blue for Social and red for Natural). A pass through this chart represents a six-month season.

Source: Author.

|  |  |
| --- | --- |
|  | A diagram of a network  Description automatically generated |

*Figure 5.* Decision making for the agents.

Source: Author.

|  |  |
| --- | --- |
| (a) | (b) |
| A group of people in a field  Description automatically generated  (c) | A video game with a green field with many people  Description automatically generated with medium confidence  (d) |

*Figure 6.* Visual outcomes. (a) Take one. (b) Take two. (c) Take three. (d) Take four.

Source: Author.



*Figure 7.* Network of Angry People. This is a network during the moment when the maximum amount of angry people was reached. You can see isolates and components. This network has an average clustering coefficient= 0.3635, 705 connected components and a density = 0.007.

Source: Author.