MANUSCRIPT

Corrupt organizations: modeling educators' misconduct with cellular automata

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Abstract Misconduct in education is a serious problem internationally. Corruption and other forms of misconduct may be modeled in large educational organizations with strong vertical and horizontal ties with the help of cellular automata. This paper offers a theoretical framework and a methodology based on cellular automata to study corruption in large educational organizations, including school districts and state and private university and college systems. Cellular automaton allows making forecasts, assessments, and predictions about the scope and scale of corruption within organizations. Starting with different cases or combinations of behavior on the workplace and working environment as initial conditions, the process of cellular automation simulates behavior of educators and results in images that depict likely future developments in educators' misconduct within educational organizations. The results of simulations reflect patterns of academic, financial, and administrative misconduct that may be found in academia.

Keywords Cellular automata \cdot Corruption \cdot Education \cdot Methodology \cdot Misconduct \cdot Modeling \cdot Organizations \cdot Peer pressure \cdot Simulation

1 Introduction

Misconduct in education is a serious problem internationally. As the education sector grows, so does the scale of misconduct. The large bureaucratic apparatus, overregulation, outdated and unclear rules, legal nihilism, labyrinth of laws and poor audit create opportunities for abuse. The blending of public sector, private firms, and

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Department of Leadership, Policy, and Organizations, Peabody College of Education, Vanderbilt University, Box 414 GPC, 230 Appleton Place, Nashville, TN 37203-5721, USA e-mail: araratos@yahoo.com personal interests of educators leads to collusion and evolvement of different forms of misconduct, especially widespread in large university systems. Educators' misconduct is not limited to embezzlement of the state funds by educational bureaucrats or collecting bribes from students by faculty members. Misconduct in education goes far beyond that and may be found in secondary and higher education sectors, in public and private sectors, in centralized and decentralized educational systems. It manifests itself in forms of bribery, embezzlement, extortion, fraud, nepotism, cronyism, favoritism, kickbacks, transgressing rules and regulations, bypass of criteria in selection and promotion, ghost teachers, cheating, plagiarism, research misconduct, data falsification, discrimination, and abuse of public property (Osipian 2007, 2008a, 2008b, 2008c, 2009a, 2009b, 2009c, 2009d, 2009e, 2009f, 2010a, 2010b, 2010c, 2010d, 2012a, 2012b, 2012c, 2012d, 2012e). In most of the instances corruption in education has a systemic character and hence can be modeled. Cellular automaton offers a promising methodology to study misconduct in education. It allows making forecasts, assessments, and predictions about the scope and scale of corruption within organizations. Cellular automata, used in sciences, may be applied to investigate corruption in large hierarchical structures of educational organizations.

2 The problem of misconduct in education

Corruption in higher education can be defined as a system of informal relations established to regulate unsanctioned access to material and nonmaterial assets through abuse of the office of public or corporate trust (Osipian 2007, p. 315). The most common forms of education corruption that may be found in different educational systems are laid down in Osipian (2007). Hallak and Poisson (2007) offer an account of education corruption throughout the world. Functions of the higher education institutions become corrupted and this corruption erodes values shared by educational organizations and the societies overall. This corruption of functions occurs among educators within rigid organizational structures, characterized by institutional rigidity. For instance, Washburn (2005) investigated the doubtful and illegal practices that occur in the relations of universities and industries and call them corporate corruption of higher education. External impact on universities done by the state may be equally important. The subject matter of education corruption as a part of misconduct in the education sector is being addressed in the scholarly work over the last five years, but the methodology of research remains very vague. Some research utilizes quantitative methodologies and survey data. Round and Rodgers (2009) study the problems of corruption in post-Soviet Ukraine's higher education sector by interviewing students in selected colleges. The research of education corruption remains on the level of conceptualization, theorization, and few methodological works. More attention is given to the nexus of political changes and educational reforms while the organizational aspects of education corruption are not well-researched. Major grounds for misconduct and corruption include the size of the system, amount of funds employed, intensity of monetary transactions, and complexity of the system. Large higher education institutions and state college and university systems employ thousands of faculty members and researchers and have budgets in billions of dollars. The literature on misconduct in education points to at least three important characteristics that are of interest for this study: the large size of educational systems and organizations where misconduct occurs, the systemic character of misconduct, and the role of peer pressure and oversight in preventing or perpetuating misconduct.

3 Literature review

Different theoretical frameworks are applied to study different forms of misconduct in large organizations. Mishra (2006) considers corruption, hierarchies and bureaucratic structures in organizations. Lui (1985, 1986) considers dynamic models of corruption and inclusion of deterrence as a factor for reducing corruption or confining it within the certain limits. Tirole (1986, 1992, 1996) brings wealth of knowledge on corruption in organizations, including hierarchies and bureaucracies, the role of collusion in organizations, collective reputations, and persistence of corruption in organizations. Carillo (2000, p. 3) points to possible collusion between supervisors and agents: "corruption can propagate within the hierarchy. We capture this recursive property of corruption by assuming that agents can share the bribe with their superiors in exchange for not being denounced." Kessler (2000) researches monitoring and collusion in hierarchies. The issue of collusion is addressed in Choi and Thum (1998), Gong (2002), Khalil and Lawarree (1993, 1995, 1996), Kofman and Lawarree (1993), Laffont and Martimort (1997), Lambert-Mogiliansky (1995), Olsen and Torsvik (1998), Strausz (1996), and Tirole (1986). These works examine collusion-proof contracts in different settings of the principal-agent frame.

Principal-agent theory, first developed in economics to study relations between the owners of the enterprises and their managers, is used to investigate corruption. The principal-agent problem in the fields of public policy, electoral process, and economics is described by Banfield (1975), Becker and Stigler (1974), Darden (2008), Klitgaard (1986), Kunicova and Rose-Ackerman (2005), Martimort (1993), Rose-Ackerman (1974, 1978, 1999), and Solnick (1998). Principals and agents are both self-interested actors, so their preferences often diverge. This agency problem not only urges a principal to monitor the agent, but also to try different mechanisms of controlling the agent's behavior. Referring to Klitgaard (1988, p. 23), Gong (2002, p. 88) states that corruption "occurs when an agent betrays the principal's interests in pursuit of his/her own or when the client corrupts the agent if he or she (client) perceives that the likely net benefits from doing so outweigh the likely net costs." Shleifer and Vishny (1993) consider vertical structures and conclude that decentralization of corruption leads to an increase in the total volume of graft collected by corrupt bureaucrats. Corruption in hierarchies is researched by Bac (1996, 1998a, 1998b, 2001), Olsen and Torsvik (1998), Osipian (2009e), and Varian (1990) in connection with the principal-agent theory. Olsen and Torsvik (1998) consider collusion in organizations within the principal-agent frame. Guriev (2004) investigates three-tier hierarchies with principal, bureaucrat, and agents. Carillo (2000) develops a four-tier hierarchical model that includes corrupt behavior. Waite and Allen (2003) follow the possible top-down and bottom-up channels of conveying benefits of corruption and resources in educational systems.

Cost-benefit analysis is used in designing cost-effective models and mechanisms of supervision. Bac (1998a, 1998b) investigates the problem of organizing three agents in a hierarchical monitoring structure and designing a corresponding incentive system to minimize the cost of implementing a target level of corruption. Bac (1996, 1998a, 1998b) combines hierarchies, cost-benefit analysis, and collusion in potentially corrupt structures and demonstrates that the possibility of collusion may prevent the implementation of anything less than full corruption. He asserts, "In relatively flat hierarchies, economies of scale in monitoring reduce implementation costs but may increase the risk of collusion." (Bac 1998a, 1998b, p. 110) Different types of hierarchies include the hierarchy where one supervisor monitors two subordinates within the supervision chain, which is shown to display in its upper part a higher risk of collusion than in its lower part. Different hierarchical structures are then contrasted with each other in order to follow the performance of each in terms of better supervision and control. Lately, methodologies normally used in sciences find their way in research of corruption, including primarily its economic aspects (Shao et al. 2007; Blanchard et al. 2005).

Lastly, one cannot leave without an acknowledgement the symposium on organizational corruption that incorporates works by Ashforth et al. (2008), Pinto et al. (2008), Lange (2008), Pfarrer et al. (2008), and Misangyi et al. (2008). The authors conceptualize different aspects of corruption in large organizational structures, such as corporations, in both micro- and macro-perspectives, searching to answer the quest for better corporate governance and more reliable and enforceable ethical standards. Osborn (1997) considers corruption as counter-culture, including population's attitudes to bribery. Organizational culture may turn to be more important than the set of formal rules and regulations imposed by the administration on the employees and by the regulatory authorities, such as the state, on the organizations. Indeed, as Ashforth et al. (2008) suggest, "a formal ethics infrastructure does not guarantee a corruption-free organization. What is required is a culture that embeds support for ethical conduct throughout its formal and informal systems." (p. 674) Helbing et al. (2011) consider self-organization and emergence in social systems by modeling the coevolution of social environments and cooperative behavior. Research of corruption with the use of cellular automata is virtually nonexistent. Wirl (1998) presents basic socio-economic typologies of bureaucratic corruption and their implications as studied through the application of cellular automata. Computational organization theory is presented in works of Carley and Prietula (1994), and Carley and Gasser (1999), among others.

There are certain specifics that make the education sector distinct from other industries, bureaucracies, and organizations in general. These specifics or distinct features and characteristics come out from the very nature of educational product—the outcome of education is not very clear and hard to measure. It is equally hard to measure the quality of education and decisions, routinely made in educational organizations. In educational organizations, educators, be it professors or administrators, set their own educational standards. These standards may vary from organization to organization and do not necessarily follow a single approved pattern. This characteristic makes them quite distinct from state bureaucracies, where standards and protocols to action are set out clearly and followed strictly. There are also disciplinary measures for non-compliance and violations, whatever minor they may be. Despite the high level of collegiality, educational organizations may be characterized by hierarchical structures, and corrupt relations in these organizations may become hierarchical as well. Osipian (2009e) analyses corruption hierarchies in higher education, presenting vertical hierarchy an institutionalized informal corrupt organization: "Vertical hierarchy anticipates both formal and informal subordination. Faculty and staff comply with formal and informal or illegal orders of top university administrators. Major characteristics of vertical hierarchy are an absolute degree of centralization and a concentration of formal and informal authority." (p. 326) Why is this important? First, in educational organizations, peer evaluation is of paramount importance. Thus, peer pressure present in professional organizations and bureaucracies often becomes the major force in educational settings. This essential feature is especially important in higher education institutions, where publications, hiring, retention, promotion, funding and other benefits are based primarily on the peer review process. Procedural defenses for vulnerable faculty members based on purely administrative rules and regulations do not work nearly as well as they would in typical bureaucracies. Second, higher-ups in an educational hierarchy can pressurize junior faculty based on artificially invented or indeed false standards, since standards are not clear. In such instances, senior faculty and administrators may serve a bad example or indeed encourage misconduct by junior faculty members. Peer to peer sharing of practices may turn into coercion.

4 Theoretical framework

As denoted by Wirl (1998, p. 203) based on works of Wolfram (1984, 1986, 1994), a cellular automaton is an iterating map F that updates at each period t the value or action of a site i, denoted a(t), depending on the neighbors actions in period (t - 1) from a fixed radius r into the set of possible states, which is discrete and of dimension k, $\{0, 1, 2, ..., k - 1\}$:

$$a_i(t) = F(a_{i-r}(t-1), a_{i-r+1}(t-1), \dots, a_{i+r}(t-1)).$$

In deterministic cellular automata, the new state of a cell is determined on the basis of its actual state and states present in the neighboring cells. In the simplest case, a onedimensional cellular automaton anticipates two possible states and a neighborhood of three cells. With two possible states and the neighborhood of three there are eight possible combinations of initial conditions and outcomes for the cell in focus. In a two-dimensional cellular automaton, cells can be positioned in hexagonal or square configurations.

In a Von Neuman neighborhood, cells are influenced by their neighbors from four sides, while in a Moore neighborhood diagonal links are also involved. Hence, a Von Neuman neighborhood consists of five cells, including the cell in focus, and a Moore neighborhood consists of nine cells. Stochastic or three-dimensional cellular automata are more complex forms than one- and two-dimensional models. In stochastic models, the transition rule allows for stochastic or probabilistic distribution. In such case the model can indicate the next state of the cell in focus based on the probability of its changing its initial state or preserving it. Stochastic cellular automaton reflects on spatial inter-specific competition of neighboring cells for the determination of the focus' cell next stage. Ideally, any large bureaucracy or professional organization, including those with complex hierarchical structures, can be decomposed to a simple linear one-period system. The resulting abstraction can be processed with cellular automata based on the set rules of functions. In some instances initial randomly distributed cells of types a and b can evolve into a homogenous state at a certain stage. In other cases, evolution will lead to a set of infinite separated simple stable or periodic structures depicting different combinations of cells a and b.

As applied to employees' behavior in complex organizations, the initial chaotic patterns of behavior can transform into periodic patterns, homogenous state, or chaotic unorganized patterns indistinguishable from the initial patterns. Periodic patterns reflect repetitive behavior of employees. Evolution leads to emergence of complex localized structures. In this case, some very complex spatial patterns may arise and reproduce over long periods of time. Such patterns may also exhibit intriguing spatial propagation despite a perfect conservation of their shape. Thus, surprisingly complex behaviors can arise from the action of randomly distributed cells with distinct patterns of behavior and result in locally concentrated processes that are not strategically directed but rather sporadic.

5 Methodology

In the simplest case, a cellular automaton consists of a line of cells or, as in our case, educators, with each cell carrying a value of zero or one. The site values evolve synchronously in discrete time steps according to the values of their nearest neighbors to indicate the effect of peer pressure and moral constraints. The analysis involves initial determination of educators who do and do not commit misconduct. The next step is to determine the period, or the single step, along the timeline. For instance, for educational financiers the period might be one financial year, while for teachers it might be one week or one academic year. The third step involves programming, or setting the rules according to which cellular automation is to progress. The rules include determinants of peer pressure and anticipated economic benefits from corruption. Further developments of the given methodology are in the two-dimensional cellular automata that can produce patterns with complicated boundaries (Packard and Wolfram 1985). Cellular automata are based on iterated functions. The process of iteration, i.e. a repetitive process, allows for an infinite number of equal steps. A limitation of the basic assumptions of the presented analysis is that automata, unlike humans, do not respond to incentives but rather act robotically. They are programmed based on the set of functions and do not act strategically.

6 Model

This paper offers the following theoretical model for application of cellular automata to misconduct in education sector and more specifically to corrupt educators. It considers educators as rational actors that calculate their expected cost and benefit of being involved in misconduct and make decisions about whether to participate in corrupt activities based on net benefits. It is assumed that net benefit from accepting a bribe or committing other possible forms of misconduct is a function of the benefits of corruption, including the size of a bribe, the risk of being exposed and prosecuted, and the social pressure from colleagues as well as personal ethics, Q = f(E, C, S). The environment in which corruption is to take place as well as the educator's personal views on corruption will be denoted as social pressure. The task is to operationalize social pressure and include it in the consideration of corrupt behavior and decision-making regarding the support of the system. We will incorporate social pressure into the initial model of corruption and compliance with the formal and informal rules that exist in the system and simulate the educator's behavior with the help of numerical examples. Social pressure includes peer pressure on the educator and his moral considerations. It is assumed that in corrupt organizations peer pressure works toward encouraging corruption. Higher peer pressure results in a higher probability for the educator to accept bribes and to comply with the current system. His/her moral considerations, however, can work in the opposite direction. Contrary to peer pressure, the educator's morality negatively impacts his willingness to accept bribes. Net social pressure is calculated by subtracting the numerical value of moral considerations from the numerical value of peer pressure. The model of decision-making based on the net benefits the educator *i* would expect from corruption is presented in the equation below:

$$Q_{i,t-1} = E_{i,t-1} + (p_{i,t-1} - m_{i,t-1}) - (d_{t-1} \times r_{t-1}), \tag{1}$$

where *i* denotes the educator, *E* is the economic benefit from being involved in corruption, *d* is the degree of punishment defined by law for a corrupt educator, *r* is the probability of being exposed, *C* is the total cost of being corrupt, *p* is the peer pressure, *m* is the moral considerations, *S* is the net social pressure, *Q* is the net benefit from corruption. All variables are taken in the period t - 1. If Q < 0, then the educator will decide not to support the current system. If Q > 0, then the educator will decide to support the current system.

Opportunity costs O of working in the education sector for period t - 1 can be equal to the educator's present legal salary L, benefits of corruption, social pressure, and risks, associated with bribery and other forms of corruption. In this case the educator is neutral to the existing system. He/she neither supports the system, nor is he/she willing to change it because his/her position in terms of income and personal wealth will likely stay unchanged. The equality can be presented as follows:

$$\mathcal{O}_{i,t-1} = L_{i,t-1} + E_{i,t-1} + (p_{i,t-1} - m_{i,t-1}) - (d_{t-1} \times r_{t-1}).$$
(2)

If O > Q + L, then the educator will decide not to support the current system. If O < Q + L, then the educator will decide to support the existing system. Peer pressure is understood as a pressure of corrupt colleagues on the educator toward corruption. Such a pressure may come from other educators within the department and the administration. Accordingly, the value of p is anticipated to always be positive. Educator makes first a decision on whether to work in the sector and, in the affirmative, whether he should take a bribe or not. The state pressure on corrupt educators is exogenous and hence is not included in the initial model. All hierarchical

Educator	ntor		Economic	Costs of corruption, risk	uption, risk		Social pressure	ssure		Net	Present	Opportunity	Decision (whether
i - 1	i	<i>i</i> + 1	benefits from corruption, <i>E</i>	Degree of punishment, <i>d</i>	Probability of being exposed, r	Total costs, <i>C</i>	Peer pressure, <i>p</i>	Moral consider- ations, <i>m</i>	Net social pressure, S	benefits, Q	legal salary, L	costs, O	to support the existing system), D
yes	yes	yes	2	4	0	0	2	0	2	4	2	3	Yes
yes	yes	ou	2	4	0.5	7	1	0	1	1	2	3	No
yes	ou	yes	2	4	0	0	2	1	1	3	2	3	Yes
yes	ou	ou	2	4	0.5	2	1	1	0	0	2	3	No
ou	yes	yes	2	4	0.5	2	1	0	1	1	2	3	No ^a
ou	yes	ou	2	4	1	4	0	0	0	-2	2	3	No
ou	ou	yes	2	4	0.5	2	1	1	0	0	2	3	No
no	ou	ou	2	4	1	4	0	1		-3	2	3	No

that the educator will likely decide not to support the system. This can be explained by some other external factors that are likely not to be in favor of supporting the system that allows corruption. Let us also explain it by some minimal transaction costs that might be incurred by the educator in order to accept bribes, embezzle, and extracts other benefits from corruption organizations use some form of internal supervision to detect corrupt behavior and wrongdoing. The educator's moral standards are assumed to be against corruption, and hence m is negative. A numerical example of defining the educator's decision of whether to support the system in exchange for the opportunities to collect bribes or commit misconduct without being punished is presented in Table 1.

7 Model simulation

Table 1 provides a numerical example for the extended model presented above (2) for the period t - 1. The assumption is made that social pressure depends on two educators who are the nearest colleagues of the educator whose decision is at stake. The educator's colleagues are denoted in the table as i - 1 and i + 1. Let us assume that the social pressure function takes the values 0 for deviating from the colleagues' behavior, 1 for conforming to one of the two colleagues, and 2 for a uniform corrupt behavior of all three educators. The values are obtained as results from the combination of peer pressure and moral considerations. Peer pressure is equal to 2 if both of the educator's colleagues are corrupt, 1 if only one of colleagues is corrupt, and 0 if both of colleagues do not accept bribes. Moral considerations are assigned values of 0 or 1, depending on whether the educator already accepts bribes. The degree of punishment for corrupt behavior is uniform for all of the possible combinations of corrupt and uncorrupt educators and has a value of 4. The probability of being exposed depends on the corruptness of the colleagues-educators. If the educator is corrupt, the probability of being exposed is equal to 0 only if both of his colleagues are corrupt. However, if the educator will accept a bribe while having both of his colleagues not involved in corrupt activities, the probability of being exposed is equal to 1. Having only one of two colleagues corrupt makes the probability of being exposed equal to 0.5. Accordingly, the value of the total cost of being corrupt varies from 0 to 2. The value of present or legal salary of the educator i is constant for all three periods, t - 1, t, and t + 1, uniform, and equal to 2. The fair market salary or the opportunity costs of the educator i is also constant for all the three periods, t - 1, t, and t + 1, uniform, and equal to 3. The value of the economic benefits from corruption is equal to 2. It is uniform for all the possible combinations. It is assumed that bribes are collected over a certain period of time. This period of time is similar to the one over which the corrupt educator bears the risk of being exposed and prosecuted. As can be seen from the numerical example, the degree of punishment is twice as high as the expected benefits from corruption. This encourages corrupt educators to seek safe harbors, such as highly corrupt environments. A good example of a safe harbor would be a department where most or all of the educators are corrupt. Let us now assume that the authorities have lowered the degree of punishment that a corrupt educator may face if accused of corruption and prosecuted. We lower the existing level of punishment of 4 down to 2. A numerical example of defining the educator's decision of whether to support the existing system in exchange for the opportunities to collect benefits of corruption without being punished is presented in Table 2.

As can be seen from Table 2, the number of cases when the educator will choose to comply with the system increased 100 percentage points, from 2 to 4. Hence, a

Truncator	Ŀ	Economic	Costs of corruption, risk	ıption, risk		Social pressure	ssure		Net	Present		Opportunity Decision (whether
i-1 i		i + 1 benefits from E	Degree of punishment, d	Probability of being exposed, r	Total costs, <i>C</i>	Peer pressure, <i>p</i>	Moral consider- ations, <i>m</i>	Net social pressure, S	benefits, \mathcal{Q}	legal salary, L	costs, O	to support the existing system), D
yes	yes ye	yes 2	2	0	0	2	0	2	4	2	3	Yes
yes	yes no	0 2	2	0.5	1	1	0	1	2	2	3	Yes
yes	no ye	yes 2	7	0	0	2	1	1	3	2	3	Yes
yes	ou ou	2	2	0.5	1	1	1	0	1	2	3	No^{a}
no	yes ye	yes 2	2	0.5	1	1	0	1	2	2	3	Yes
no	yes no	2	2	1	2	0	0	0	0	2	3	No
no	no ye	yes 2	2	0.5	1	1	1	0	1	2	3	No
no	ou ou	0 2	2	1	2	0	1			2	3	No
no no ^a Similar	no y no ni to the J	eriod	2 2 ie case in the nun	0.5 1 nerical example	1 2 e in the p	$\frac{1}{0}$	1 1 3pportunity 6	0 -1 sosts of the ed	$\frac{1}{-1}$	2 2 ; equal to th	3 3 he sum of his	pr

Table 2 A numerical example of defining the educational employee's decision of whether to support the system, based on such considerations, as total benefit, costs, and

voluntary reduction of the degree of punishment from 4 in period t - 1 down to 2 in period t leads to a significant increase in the number of cases in which the educator will support the system.

Despite the significant increase in the number of cases when the educator will support the existing system in period t, it constitutes only half of all possible cases. This is not sufficient for the system that wants to sustain itself. The system cannot afford an increase in the salaries it pays to college professors due to budget constraints. Nor can it facilitate an increase in the total sum of benefits educators generate from corruption. The size of bribes and the total scale and scope of bribery and other forms of corruption in education, as well as in other sectors of the economy, are mostly determined by the market forces, including consumer demand and clientele base, not by the state. Further proliferation of the corruption and compliance policy is needed. Therefore, as follows from (1) and (2), the authorities are interested in the reduction of the total cost of being involved in corruption for each educator. This can be done easily since the punishment mechanism is administered by the state. While the state cannot regulate the risk of exposure r, it can regulate the degree of punishment d. The degree of punishment consists of the probability of being prosecuted and sentenced and the level of punishment chosen by the state in regard to the corrupt educator. While formally the degree of punishment may be high, the actual degree of punishment d may be relatively low, based on the low rate of prosecution. Furthermore, prosecution itself is a threat only for those who choose not to comply with the authorities' demands.

Let us assume that the authorities have lowered the degree of punishment that a corrupt educator may face if accused of corruption and prosecuted. We reduce the existing level of punishment of 2 in period t down to 1 in period t + 1. A numerical example of defining the educator's decision of whether to support the system in exchange for the opportunities to collect bribes without being punished is presented in Table 3. The number of cases when the educator will chose to comply with the system's demands increased 50 percentage points, from 4 to 6. Hence, a further voluntary reduction of the degree of punishment from 2 in period t down to 1 lead to a significant increase in the number of cases in which the educator will opt for supporting the system. Probability of being exposed may be a function of peer pressure. Accordingly, an increase in peer pressure may lead to a decrease in the probability of being exposed and, hence, to a further decrease in the total cost of being involved in corrupt activities. This will lead to an even higher probability of the educator being in support of the existing system.

8 Results

The results of cellular automation simulation, including those obtained after analyzing the large educational organizations, are best seen as graphic depictions. They might be simple yet reliable assessments of the future developments that reflect the scale and the scope of educational misconduct. Wirl says that "Although cellular automata are very simple, deterministic machines and thus crude approximations of real, economic situations, they are capable of describing self organization and complex patterns (of corruption)." (Wirl 1998, p. 199) The images, both black and white

Educator			Economic	Costs of corruption, risk	uption, risk		Social pressure	ssure		Net	Present	Opportunity	Decision (whether
i - 1	i – 1 i	<i>i</i> + 1	i+1 benefits from E	Degree of punishment, d	Probability of being exposed, <i>r</i>	Total costs, <i>C</i>	Peer pressure, <i>p</i>	Moral consider- ations, <i>m</i>	Net social pressure, S	benefits, Q	legal salary, L	costs, O	to support the existing system), D
yes	yes	yes	2	1	0	0	2	0	2	4	2	e,	Yes
yes	yes	ou	2	1	0.5	0.5	1	0	1	2.5	2	3	Yes
yes	ou	yes	2	1	0	0	2	1	1	3	2	3	Yes
yes	ou	ou	2	1	0.5	0.5	1	1	0	1.5	2	3	Yes
ou	yes	yes	2	1	0.5	0.5	1	0	1	2.5	2	3	Yes
ou	yes	ou	2	1	1	2	0	0	0	1	2	3	No^{a}
no	ou	yes	2	1	0.5	0.5	1	1	0	1.5	2	3	Yes
no	ou	ou	2	1	1	7	0	1		0	2	3	No

Table 3 A numerical example of defining the educational employee's decision of whether to support the system, based on such considerations, as total benefit, costs, and social pressure (period $t \neq 1$).

opposition to the existing system

and in color, depending upon the initial characteristics of the cells and the authors' determination, allow for visual examination of future patterns of misconduct. The structures with the clear aisles or sporadic distribution of corrupt educators point to-ward particular educators who are likely to commit misconduct in the future. Most interestingly, the predictions point to those members of large organizations who are most likely to be involved in misconduct after a certain period of time and yet who at the present may even be unaware of this.

We present three simulations based on distinct functions of deterministic patterns of behavior. The images appear structuralistic in nature, with dispersed triangles of different sizes, often localized in groups, with diffused and randomly distributed single cells. In all of the images generated below, black color identifies a corrupt educator, while white color identifies a non-corrupt educator. Two neighbors, one on the left and one on the right, influence their neighbor in the middle. We focus on the educator in the middle. For each function, we use 1000 educators in a one-year, i.e. 365-day period, where each cell represents a given educator in a given day.

We present three functions. Each of the functions reflects a certain balance of powers and combination of factors, including central authorities, educators, pay rates, risk of exposure, degree of punishment, and peer pressure. Based on the significance of these initial factors in each of the three cases, we formulate certain dependencies expressed as functions 1, 2, and 3.

Function 1. (Rule 18). Let us assume that: (1) three corrupt educators grouped together cause the authorities to initiate an investigation; accordingly, the risk of punishment for being involved in misconduct increases, and as a result the educator refuses to participate in corruption. Hence, having two corrupt neighbors in period t-1causes the educator to become non-corrupt in period t; (2) having one non-corrupt neighbor causes the corrupt educator to become non-corrupt in period t, if he was corrupt in period t - 1; (3) having two corrupt educators-neighbors causes the educator to remain non-corrupt in period t, because he/she reasonably expects that his/her neighbors will remain corrupt in period t and that three corrupt educators will cause the authorities to initiate an investigation. The risk will go up and the educator will have to refuse corruption; (4) having two non-corrupt neighbors causes the corrupt educator to become non-corrupt, since peer pressure in this case pushes him/her toward non-corruption. In addition, the risk of being exposed by non-corrupt peers is higher; (5) finally, having two non-corrupt neighbors in period t - 1 causes the noncorrupt educator to remain non-corrupt in period t. The results of cellular automaton for the function 1 are presented in Figs. 1, 2, 3, 4.

Function 2. (Rule 126). Let us assume that: (1) three corrupt educators grouped together cause the authorities to initiate an investigation; accordingly, the risk of punishment for being involved in misconduct increases, and as a result the educator refuses to participate in corruption. Hence, having two corrupt neighbors in period t - 1 causes the educator to become non-corrupt in period t; (2) having one corrupt neighbor allows the corrupt educator to remain corrupt in period t, if he was corrupt in period t - 1; (3) having one corrupt neighbor in period t; (4) having two corrupt educators-neighbors in period t; (5) having two non-corrupt neighbors allows the corrupt neighbors allows the corrupt educator to become corrupt educator to period t - 1 allows non-corrupt educator to remain corrupt in period t; (5) having two non-corrupt neighbors allows the corrupt educator to remain corrupt in period t; (5) having two non-corrupt neighbors allows the corrupt educator to remain corrupt in period t; (5) having two non-corrupt neighbors allows the corrupt educator to remain corrupt in period t; (5) having two non-corrupt neighbors allows the corrupt educator to remain corrupt.

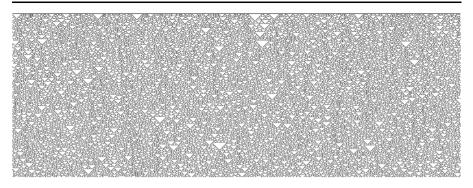
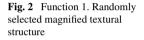
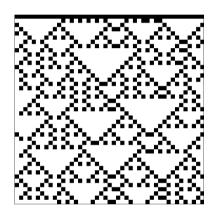


Fig. 1 Function 1. Cellular automaton for 1000 educators in a 365-day period, with corrupt educators being distributed randomly in day one





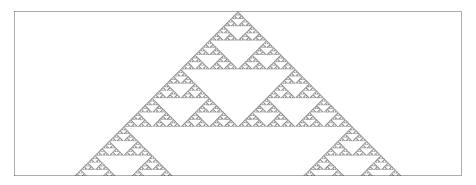


Fig. 3 Function 1. Cellular automaton for 1000 educators in a 365-day period, with one corrupt educator initially in day one

pressure in this case is weaker and does not push him/her toward non-corruption. In addition, the risk of being exposed by non-corrupt peers is lower; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain

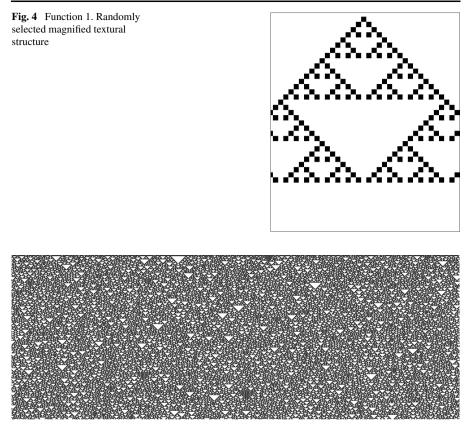


Fig. 5 Function 2. Cellular automaton for 1000 educators in a 365-day period, with corrupt educators being distributed randomly in day one

non-corrupt in period t. The results of cellular automaton for the function 2 are presented in Figs. 5, 6, 7, 8.

Function 3. (Rule 86). Let us assume that: (1) three corrupt educators grouped together cause the authorities to initiate an investigation; accordingly, the risk of punishment for being involved in misconduct increases, and as a result the educator refuses to participate in corruption. Hence, having two corrupt neighbors in period t - 1 causes the educator to become non-corrupt in period t; (2) having one corrupt neighbor causes the non-corrupt educator to become corrupt in period t; (4) having two corrupt educators to remain non-corrupt in period t, because he/she reasonably expects that his/her neighbors will remain corrupt in period t and that three corrupt neighbors allows the corrupt educator to remain corrupt in period t - 1 causes the non-corrupt; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain non-corrupt neighbors in period t - 1 causes the non-corrupt; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt in period t - 1 causes the non-corrupt in period t - 1 causes the non-corrupt; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain non-corrupt neighbors in period t - 1 causes the non-corrupt; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain non-corrupt in period t - 1 causes the non-corrupt in period t - 1 causes the non-corrupt in period t - 1 causes the non-corrupt; (6) finally, having two non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain non-corrupt in period t - 1 causes the non-corrupt educator to remain non-corrupt in period t - 1 causes the non-corrupt educator to remain non-corrupt neighbors in period t - 1 causes the non-corrupt educator to remain non-corrupt neig

Fig. 6 Function 2. Randomly selected magnified textural structure

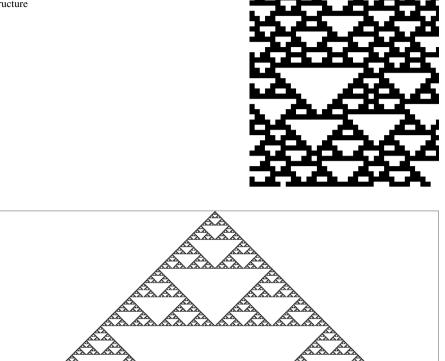
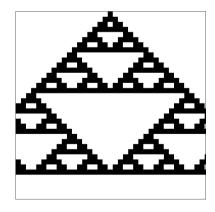


Fig. 7 Function 2. Cellular automaton for 1000 educators in a 365-day period, with one corrupt educator initially in day one

Fig. 8 Function 2. Randomly selected magnified textural structure



Functions 1, 2, and 3, depicted on the images, do not necessarily correspond with the numerical examples we offered earlier. But in the essence, lesser peer pressure to be non-corrupt and the risks associated with participation in corrupt activities become definitive in educators' behavior in both numerical simulations and graphic

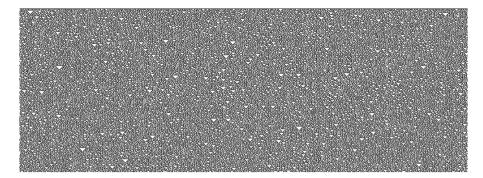
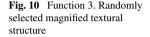
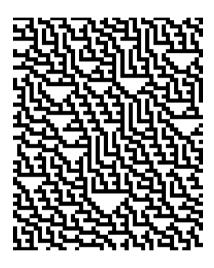


Fig. 9 Function 3. Cellular automaton for 1000 educators in a 365-day period, with corrupt educators being distributed randomly in day one





representations. According to Function 1, the educator is unlikely to be encouraged to participate in misconduct in most of the instances. As a result, the structure of the cellular automaton for 1000 educators in a 365-day period, with corrupt educators being distributed randomly in day one and with one corrupt educator initially in day one, depicted in Figs. 1 and 3, respectively, is of a lesser density than that of Function 2. Cellular automaton based on Function 2 appears to have somewhat similar structure, but is clearly denser. This means that higher peer pressure to become corrupt and lesser risk of prosecution make the number of instances of having corrupt educators is much higher.

Finally, as depicted in Fig. 9, cellular automaton based on Function 3 is less chaotic and has a more structured appearance, than do cellular automata based on Functions 1 and 2. Figure 11 presents a quite astonishing pattern of distribution of educators' misconduct that starts from a single corrupt educator in day 1 and by the end of the year there are already a few hundred corrupt educators with a perspective of further proliferation until the margins are reached. The triangle that reflects the area of misconduct spread in the educational organization has a much higher density

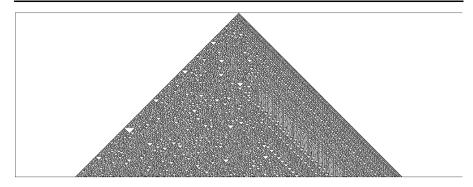
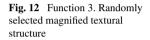
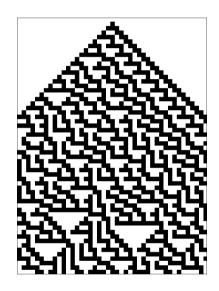


Fig. 11 Function 3. Cellular automaton for 1000 educators in a 365-day period, with one corrupt educator initially in day one





than the pyramidal structures in Figs. 3 and 7. Equally interesting is that there is a clearly visible asymmetry in the way the cellular automaton progression is structured. The right side of the triangle and its center is structured along horizontal and vertical lines, while the left side of the triangle is grouped more along the diagonal lines directed from the center parallel to the left lateral position.

9 Inferences, organizational implications, and policy recommendations

Social pressure and peer pressure in particular appear to be definitive when it comes to corruption in large educational organizations, and is of paramount importance in higher education institutions. Lesser peer pressure to be non-corrupt and the risks associated with participation in corruption become definitive in educators' corrupt behavior. To reverse the trend on corruption, higher ethical and moral standards may be needed. The state may be interested in maintaining a certain level of corruption in

autonomy, which may be found throughout the world.

educational organizations in order to advance the agenda of a political regime. Corrupted organizations and educators are easy to manipulate. In such instances, more internal control may be suggested instead of less effective external governmental control. This suggestion is in line with aspirations for higher degrees of the university

Social and work climate in educational organizations is especially important, because all organizational decisions, including academic appointments and promotions, are based on the peer review process, including both internal and external peer review. Peer pressure may weed out corruption by forcing the "bad apple" educators to act ethically or eliminating them by pushing them outside the organization or it may well chase good educators away while letting corrupt educators to "infect" their colleagues with the virus of corruption, causing corruption to proliferate. Some organizational environments have a very welcoming climate for corrupt activities. As depicted in Fig. 11 of Function 3, just one corrupt educator can turn a large educational organization in a hotbed of corruption in less than two years.

Each educational organization has the duty to cultivate intellectual integrity as a combination of academic integrity and research and scholarly integrity. This duty rests primarily with the organization rather than the state. Each higher educational organization, taken as a kind of semi-autonomous educational enclave within the national educational system, is used as a unit of analysis. Naturally, it has its own subunits and subdivisions, such as departments. Daily actions of educators shape a collective reputation of each department and sum up to the reputation of the entire organization. As simulations show, an increased internal control alone is not going to solve the problem of corruption. In addition, internal control is costly, although it may be less costly than externally imposed governmental control. There is a need for more incentives for a good behavior, including incentives of not only individual character, but of collective character as well.

In order to better link the recommendations with the simulation results, derived from the models, we will try to address the question about the immunity of organizations of higher learning in regard to corrupt educators spreading the virus of corruption. Simply put, which types of organizations are vulnerable to intrusion of one corrupt individual, and which kind of organizations resist better to corruption? In better protected organizations, the risk of exposure actually materializes in proper reaction of the organizational authorities and external supervisors, such as state agencies. According to first and second functions, the risk of exposure increases with the grouping of corrupt educators. As under the military rule in some dictatorial regimes, the "do not gather three or more" rule applies to corrupt educators, in which way their "freedom of assembly" becomes violated. Such is the negative effect of peer pressure. Under the bad apple approach, a corrupt educator, even if implanted randomly into an organization, will eventually be identified, exposed, and punished or dismissed. However, according to the third function, if a corrupt educator gets seeded in a welcoming environment, susceptible to corruption, corrupt practices may well proliferate among previously uncorrupt educators, instead of this bad apple educator being weeded out from the institution of higher learning. So the answer to the question is that the organization's immunity or the ability to resist corruption depends not as much on the structural specifics but organizational culture supported by oversight

and internal stimuli. Indeed, not educational organization may be immune to corrupt educators, but some manage to not let corruption proliferate at a high rate.

If just one corrupt educator can turn a large educational organization into a hotbed of corruption, how can the organization match this risk and meet it with proper anticorruption tactics? Which types of organizations can safeguard from corrupt educators coming in and infiltrating higher education institutions? Mere vigilance and risk aversion strategies in employing and promoting academics may not help much, for it is the environment that often plays a significant or even decisive role in corrupting academics. The human factor, so persistently emphasized in the former socialist states, points to the robotic versus human actors' controversy. However, this controversy does not fundamentally undermine the validity of the inferences one makes for the model, because the model simulations rely on incentives, and human actors, especially those teaching and administering institutions of higher learning, are quite rational ones. The information produced by the simulation is actionable, but the inferences drawn from it may of course be challenged. Such constraint with respect to this perspective seems to be natural, and nevertheless a cellular automation based simulations could prove useful in modeling educators' illicit behavior in a wide variety of different higher educational environments. Different functions of interactions and mutual interdependencies among educators introduced in this work depict variations in simple organizational structures. Peer pressure in research oriented universities may be more significant and at the same time less structured and administratively controlled, than it is in liberal art colleges or so-called teaching schools. For different educational systems and/or types of educational organizations different functions apply. Accordingly, policy recommendations on how to curb corruption may differ. In the former socialist systems, where faculty members are routinely underpaid, higher legal income may be a partial solution to the problem of endemic corruption. For the US, better internal control may work better in combination with the less corrupt governmental oversight.

The overreliance on the governmental control and/or salary increases may prove a wrong way to tackle corruption in large educational organizations. The state may intentionally lower the risk of punishment or such risk may be very low due to the high degree of corruptness of a given educator's peers. In both cases, corruption will simply proliferate to even higher levels. In the case of higher salaries, this target is simply hard to achieve due to the continuing commercialization of the education sectors throughout the world. Salaries of educators are less and less dependent on the state and more and more dependent on the market forces. In a competitive environment of the market, educators representing their organizations contribute to and rely on reputation of their educational institution. This reputation may be good or bad, and is a result of balancing positive and negative signals about this organization present on the market over a long time span. Thus, collective reputations backed by the peer pressure to act ethically are the key in the fight against corruption in educational organizations.

Cellular automaton may prove to be a more effective and cost-efficient methodology than estimation of systems of partial differential equations. Some of the aspects of organizational corrupt structures may be studied along the lines of computational organization theory which uses computational and mathematical methods to study organizations, formulates models, and develops tools and procedures to validate organizational models. Eventually, this methodology will be used to improve educational organizations through an increase in their organizational effectiveness and efficiency and a reduction and future prevention of misconduct.

10 Concluding remarks

This paper presents cellular automation, a relatively new methodology to study misconduct in large educational organizations, and uses simulation to model the behavior of educators, including factors that influence their decision making. This methodology may be used beneficially for future research in organizations and corrupt hierarchies, including school districts and higher education institutions and make valid and credible forecasts. Cellular automaton is not universal, as any other methodology. Unlike humans, it acts robotically. Also, diversity of organizational structures and informal networks in higher education institutions as well as a wide variety of forms of academic misconduct around the world posits a challenge to this approach. Nevertheless, cellular automaton based simulations can be used to model a wide variety of different environments and patterns of development, from corrupt practices among faculty in Tbilisi State University in the country of Georgia to education policy adoption strategies of the state of Georgia in the US, and from distinct modes of research misconduct in large research universities and think tanks to opportunistic behavior of education bureaucrats and faculty members in large higher education systems.

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