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# Continuous opinions and discrete actions in social networks: a multi-agent system approach

N. R. Chowdhury, I.-C. Morărescu, S. Martin, S. Srikant

**Abstract**—This paper proposes and analyzes a novel multi-agent opinion dynamics model in which agents have access to actions which are quantized version of the opinions of their neighbors. The model produces different behaviors observed in social networks such as disensus, clustering, oscillations, opinion propagation, even when the communication network is connected. The main results of the paper provides the characterization of preservation and diffusion of actions under general communication topologies. A complete analysis allowing the opinion forecasting is given in the particular cases of complete and ring communication graphs. Numerical examples illustrate the main features of this model.

## I. INTRODUCTION

The analysis of social networks received an increasing interest during the last decade. This is certainly related with the increasing use of Facebook, Twitter, LinkedIn and other on-line platforms allowing to get information about social networks. Multi-agent systems have been shown to efficiently model how opinion dynamics occur as a result of social interactions [1]. The existing models can be split in two main classes: those considering that opinions can evolve in a discrete set and those considering a continuous set of values that can be taken by each agent. The models in first class come from statistical physics and the most employed are the Ising [2], voter [3] and Sznajd [4] models. When the opinions are not constrained to a discrete set, we can find in the literature two popular models: the Deffuant [5] and the Hegselmann-Krause [6] models. They are usually known as bounded confidence models since they depend on one parameter characterizing the fact that one agent takes into account the opinion of another only if their opinions are close enough. The bounded confidence models above do not guarantee consensus and instead several local agreements can be reached. The Hegselmann-Krause model has been adapted in [7] to a model of opinion dynamics with decaying confidence.

In order to more accurately describe the opinion dynamics and to recover more realistic behaviors, a mix of continuous opinion with discrete actions (CODA) was proposed in [8]. This model reflects the fact that even if we often face binary choices or actions, the opinion behind evolves in a continuous space of values. For instance we may think that car A is 70% more appropriate for our use than car B. However, our action will be 100%

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buy car A. Moreover, our neighbors often only see our action without any access to our opinion. Similar idea was employed in [9] where it is studied the emergence of consensus under quantized all-to-all communication. In this paper the authors assume constant interaction weights and quantized information on the opinion of all the other individuals belonging to a given social network. In [10], a system with quantized communication and general communication topologies is studied. There, the author focuses on consensus (up to the quantizer precision) while in the present paper, we precisely focus on dynamics such as dissensus which occur within the quantizer precision. As explained above, this is of particular relevance for the social sciences.

Whatever is the model employed to describe the opinion dynamics, many studies focus on the emergence of consensus in social networks [11], [12], [5], [13]. Nevertheless, this behavior is rarely observed in real large social networks. The present study provides a possible explanation for dissensus even though the communication graph is connected.

Conceptually the model proposed in this paper is close to the one in [9] with the difference that we are considering state-dependent interaction weights instead of constant ones. Beside the heterogeneity introduced in the model by the state-dependent interaction weights we are also dealing with more general interaction topologies and we are not trying to characterize consensus. Instead we highlight that extremist individuals are less influenceable, that several equilibria can be reached and one can also have oscillatory behaviors in the network. From behavioral view-point our model is close to the one in [8] (this similarity is highlighted in Section II). The difference here is that we are able to analytically study this model and go beyond simulations and their interpretation.

The main contributions of this work are threefold. **(i)** We propose a consensus-like dynamics that approaches the dynamics described in [8]. This CODA model is given by a quantized consensus system with state-dependent interaction weights. **(ii)** We describe the possible equilibria of the proposed model and depict the main properties characterizing the opinions dynamics. Precisely, we provide a criterion to detect stabilization of the actions of a group of agents and to predict the propagation of this action throughout the network. Our criterion depends on the initial conditions and interaction topology only. **(iii)** We completely analyze some particular interaction topologies such as: the complete graph and the ring graph.

The rest of the paper is organized as follows. Section II formulates the problem and illustrates that our model is close to the one proposed in [8]. In Section III we show when the quantization effect is removed *i.e.* in the context of continuous opinions with continuous actions (COCA), that

consensus is always achieved provided that the interaction graph is connected. The main features of our CODA model under general interaction topologies are derived in Section IV. Precisely we characterize the preservation and the propagation of actions inside a group and outside it, respectively. Some particular interaction topologies such as: the complete graph and the ring graph are analyzed in V.

## II. PROBLEM FORMULATION AND PRELIMINARIES

We consider a network of  $n$  agents denoted by  $\mathcal{V} = \{1, \dots, n\}$ . The interaction topology between agents is described by a fixed graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  that can be directed or not. Let  $N_i$  be the set of agents that influence  $i$  according to the graph  $\mathcal{G}$  (i.e.  $j \in N_i \Leftrightarrow (j, i) \in \mathcal{E}$ ) and  $n_i$  be the cardinality of  $N_i$ . Initially, agent  $i \in \mathcal{V}$  has a given opinion  $p_i(0) = p_i^0 \in (0, 1)$  and this opinion evolves according to a discrete-time protocol. Let  $p_i(k)$  be the opinion of agent  $i \in \mathcal{V}$  at time  $k$ . Assume that  $\forall i \in \mathcal{V}$ ,  $p_i(0) \neq \frac{1}{2}$  and introduce the action value  $q_i(k) \in \{0, 1\}$  as a quantized version of  $p_i(k)$  defined by:

$$q_i(k) = \begin{cases} 0 & \text{if } (p_i(k) < \frac{1}{2}), \\ 0 & \text{if } (p_i(k) = \frac{1}{2} \text{ and } p_i(k-1) < \frac{1}{2}), \\ 1 & \text{otherwise.} \end{cases}$$

Two distinct situations are considered in the following.

**COCA model:** each agent has access to the opinion of its neighbors. In this case, our model simply writes as a consensus protocol with state-dependent interaction weights. Precisely, the opinion of an agent  $i \in \mathcal{V}$  updates according to the following rule

$$p_i(k+1) = p_i(k) + \frac{p_i(k)(1-p_i(k))}{n_i} \sum_{j \in N_i} (p_j(k) - p_i(k)). \quad (1)$$

Denoting by  $p(\cdot) = (p_1(\cdot), \dots, p_n(\cdot))$  the vector that collects the opinions of all agents, the collective opinion dynamics is given by:

$$p(k+1) = (I_n + A(p(k)))p(k), \quad (2)$$

where  $A(p(k))$  is the opposite of the Laplacian matrix associated to adjacency matrix  $(a_{ij}(k))_{ij}$  with off-diagonal weights  $a_{ij}(k) = p_i(k)(1-p_i(k))/n_i$  if  $j \in N_i$  and  $a_{ij}(k) = 0$  otherwise.

*Remark 1:* We assume that  $p_i^0$  belongs to  $(0, 1)$  which is a normalized version of  $\mathbb{R}$ . Doing so, the matrix  $I_n + A(p(k))$  is row stochastic and for all  $k \in \mathbb{N}$  one has  $p(k) \in (0, 1)$ .

**CODA model:** provides the main model under study in this paper. This model assumes that each individual has access only to the action of its neighbors. The opinion of agent  $i \in \mathcal{V}$  in this case updates according to the following rule:

$$p_i(k+1) = p_i(k) + \frac{p_i(k)(1-p_i(k))}{n_i} \sum_{j \in N_i} (q_j(k) - p_i(k)). \quad (3)$$

*Remark 2:* For an agent  $i \in \mathcal{V}$ , both COCA and CODA models propose interaction weights depending only on the opinion  $p_i$ .

We can emphasize a natural partition of  $\mathcal{V}$  in two subsets  $N^-(k) = \{i \in \mathcal{V} \mid q_i(k) = 0\}$  and  $N^+(k) = \{i \in \mathcal{V} \mid q_i(k) = 1\}$ . The main objective of this paper is to study how these sets evolve in time and what is the behavior of the opinions  $p_i(k)$  inside these sets. Throughout the paper we denote by  $n^-(k)$  and  $n^+(k)$  the cardinality of  $N^-(k)$  and  $N^+(k)$ , respectively. Similarly, for an agent  $i$  we denote by  $N_i^-(k) = N_i \cap N^-(k)$  and  $N_i^+(k) = N_i \cap N^+(k)$  and by  $n_i^-(k)$  and  $n_i^+(k)$  the cardinalities of these sets.

The CODA model (3) was inspired by Martins' model [8] which was formulated in terms of the following bayesian update. Let us denote by  $\tilde{p}_i(k)$  the opinion of agent  $i \in \mathcal{V}$  at time  $k \in \mathbb{N}$  when using Martins' model [8]. The updates of this opinion follows the rule described below. When agent  $i$  is influenced by agent  $j$ ,

$$\frac{\tilde{p}_i(k+1)}{1-\tilde{p}_i(k+1)} = \frac{\tilde{p}_i(k)}{1-\tilde{p}_i(k)} \cdot \frac{\alpha}{1-\alpha} \text{ if } q_j(k) = 1, \quad (4)$$

and where  $\alpha$  is replaced by  $1-\alpha$  if  $q_j(k) = 0$ . The constant  $\alpha \geq 0.5$  is a model parameter linked to the amplitude of opinion change as a result of interactions. This parameter does not appear in our model. The study [8] is based on numerical experiments and do not present a theoretical analysis. The simulations found in [8] where obtained with  $\alpha = 0.7$ . The simulations appear to be qualitatively close to the ones resulting from our update (3) (see Section VI). To understand this fact, one can show that for  $\alpha = 2/3 \approx 0.7$ , and  $q_j(k) = 1$ , update (3) and (4) are equivalent for small  $p_i(k)$  values. Similarly, if  $q_j(k) = 0$  the two updates are equivalent for  $p_i(k)$  values close to 1. For other  $p_i(k)$  values,  $p_i(k+1)$  still remains close to  $\tilde{p}_i(k+1)$ . These facts are illustrated in Figure 1. As a consequence, our CODA model (3) can be seen as a consensus-type version of Martins' model. One advantage of our model is to explicitly explains why extremist agents with opinion close to 0 or 1 hardly change their actions. This is due to the weight  $p_i(1-p_i)$  in update (3).

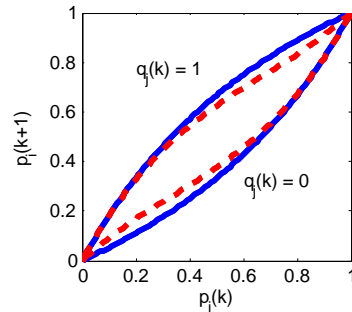


Fig. 1. Comparison of the CODA model (3) and the model by Martins [8]. The two top curves display the update (3) and (4) when the influencing agent has action  $q_j = 1$ . The two bottom curves display the update (3) and (4) when the influencing agent has action  $q_j = 0$ . Discontinuous red lines correspond to update (3) while strait blue lines correspond to update (4).

## III. ANALYSIS OF THE COCA MODEL

When agents have a direct access to opinions without quantification, opinions tend to a consensus at exponential speed as is shown in the present section. This is no longer the case when

agents only have access the quantized actions instead of opinions of their neighbors, as will be shown in Section IV. It is noteworthy that an opinion stays constant according to dynamics (1) if it is initialized at the value 0 or 1. Therefore, by the following assumption we exclude these extreme cases from the analysis.

*Assumption 1:* There exists a *strictly* positive constant  $\epsilon \in (0, \frac{1}{2})$  such that for all  $i \in \mathcal{V}$ , one has  $p_i^0 \in [\epsilon, 1 - \epsilon]$ .

For a subset of agents,  $\mathcal{V}' \subseteq \mathcal{V}$  and for all  $k \in \mathbb{N}$ , let us define  $m_{\mathcal{V}'}(k) := \min_{j \in \mathcal{V}'} p_j(k)$  and  $M_{\mathcal{V}'}(k) := \max_{j \in \mathcal{V}'} p_j(k)$ . The following result can be easily proven by induction.

*Lemma 1:* Let  $\mathcal{V}' \subseteq \mathcal{V}$  such that  $\mathcal{V}'$  is isolated *i.e.* for all  $i \in \mathcal{V}'$ , for all  $j \in \mathcal{V} \setminus \mathcal{V}'$ ,  $(j, i) \notin \mathcal{E}$ . Then,

- 1)  $M_{\mathcal{V}'}(k+1) \leq M_{\mathcal{V}'}(k)$
- 2)  $m_{\mathcal{V}'}(k+1) \geq m_{\mathcal{V}'}(k)$

Let us assume for the rest of this section that the graph is strongly connected.

Applying Lemma 1 with  $\mathcal{V}' = \mathcal{V}$  one obtains that sequences  $m_{\mathcal{V}}(k)$  and  $M_{\mathcal{V}}(k)$  are both monotonic and bounded, thus convergent. The interaction weight from the agent  $j \in N_i$  to the agent  $i$  at time  $k$  is defined by  $a_{ij}(k) := \frac{p_i(k)(1-p_i(k))}{n_i}$ . Taking into account that  $p_i^0 \in [\epsilon, 1 - \epsilon]$ , the behavior of the function  $x \mapsto x(1-x)$  on  $(0, 1)$  and Lemma 1, one deduces that  $a_{ij}(k) \in [\frac{\epsilon(1-\epsilon)}{n_i}, \frac{1}{4n_i}]$  for all  $k \in \mathbb{N}$  and  $j \in N_i$ . Moreover, for all  $i \in \mathcal{V}$  one has  $1 \leq n_i \leq n-1$  yielding

$$a_{ij}(k) \in \left[ \frac{\epsilon(1-\epsilon)}{n-1}, \frac{1}{4} \right], \quad \forall i \in \mathcal{V}, \forall j \in N_i. \quad (5)$$

Since  $a_{ii} = 1 - \sum_{j \in N_i} a_{ij}$ , straightforward computation shows that

$$a_{ii}(k) \in \left[ \frac{3}{4}, 1 - \epsilon(1-\epsilon) \right], \quad \forall i \in \mathcal{V}. \quad (6)$$

Equation (5) and (6) shows that Assumption 1 in [14] holds true. Moreover, in this paper we are dealing with a fixed (strongly) connected graph meaning that Assumption 2 and 3 in [14] also hold. Therefore the following result is a consequence of Theorems 1 and Lemma 1 in [14].

*Proposition 2:* As far as the graph  $\mathcal{G}$  is (strongly) connected and Assumption 1 is satisfied, the update rule (1) guarantees asymptotic consensus of all opinions. Furthermore, there exist two constants  $\beta \in (0, 1)$  and  $T \in \mathbb{N}$  such that  $\forall k \in \mathbb{N}$

$$M_{\mathcal{V}}(k+T) - m_{\mathcal{V}}(k+T) \leq \beta(M_{\mathcal{V}}(k) - m_{\mathcal{V}}(k)).$$

#### IV. ANALYSIS OF THE CODA MODEL

For the remaining part of this work we consider the update rule (3). In this section we derive the set of possible equilibrium points and we give the conditions guaranteeing that the action  $q_i$  of the agent  $i \in \mathcal{V}$  is preserved/changed over time. Throughout the rest of the paper we assume that  $\forall i \in \mathcal{V}$ ,  $p_i(0) \in (0, 1)$ .

#### A. Characterization of equilibria

Let us define the following finite set of rational numbers:

$$\mathcal{S} = \left\{ \frac{k}{m} \mid k, m \in \mathbb{N}, k \leq m \leq n-1 \right\} \subset \mathbb{Q}. \quad (7)$$

The main result of this section states that the possible equilibrium points of the opinions belong to  $\mathcal{S}$ . Let us first introduce an instrumental result. For the rest of the paper we use the notation  $r_i(k) := \frac{n_i^+(k)}{n_i}$ .

*Lemma 3:* Let  $i \in \mathcal{V}$ ,  $p_i(0) \in (0, 1)$ . Then, for all  $k \in \mathbb{N}$ , one of the following relations holds

$$p_i(k) < p_i(k+1) < r_i(k), \quad (8)$$

$$p_i(k) > p_i(k+1) > r_i(k), \quad (9)$$

or

$$p_i(k) = p_i(k+1) = r_i(k). \quad (10)$$

Moreover,  $\forall k \in \mathbb{N}$ ,  $p_i(k) \in (0, 1)$ .

*Proposition 4:* Let  $i \in \mathcal{V}$  and suppose Assumption 1 holds. If  $(n_i^+(k))_{k \geq 0}$  (and thus  $(n_i^-(k))_{k \geq 0}$ ) is stationary, then the sequence of opinions  $(p_i(k))_{k \geq 0}$  converges and has the following limit :

$$p_i^* = \lim_{k \rightarrow \infty} p_i(k) = \frac{\lim_{k \rightarrow \infty} n_i^+(k)}{n_i} \in \mathcal{S}.$$

Conversely, if the sequence of opinions  $(p_i(k))_{k \geq 0}$  converges towards a non-extreme value  $p_i^* \in (0, 1)$ , then the sequence  $(n_i^+(k))_{k \geq 0}$  is stationary and the equality between the limits still holds.

*Proof:* If the sequence  $(n_i^+(k))_{k \geq 0}$  is stationary, one gets that  $n_i^+(k)$  is constant for  $k$  bigger than or equal to a fixed integer  $k^*$ . Let us denote by  $\rho_i^*$  the value of  $r_i(k)$  for  $k \geq k^*$ . According to Assumption 1,  $p_i(0) \in (0, 1)$ , so that Lemma 3 applies. By induction one has either

$$\forall k \geq k^*, p_i(k) \leq p_i(k+1) \leq \rho_i^* \quad (11)$$

or

$$\forall k \geq k^*, p_i(k) \geq p_i(k+1) \geq \rho_i^*. \quad (12)$$

In the first case  $(p_i(k))_{k \geq 0}$  is increasing and upper-bounded, in the second  $(p_i(k))_{k \geq 0}$  is decreasing and lower-bounded. Thus  $(p_i(k))_{k \geq 0}$  converges. Denote  $p_i^*$  its limit. If  $p_i^* = 1$ , equation (11) must be satisfied and  $\rho_i^* = 1$ . If  $p_i^* = 0$ , equation (12) must be satisfied and  $\rho_i^* = 0$ . Finally, if  $p_i^* \in (0, 1)$ , one can take the limit of equation (3) to show that  $p_i^* = \rho_i^*$ .

Reversely, assume that the sequence of opinions  $(p_i(k))_{k \geq 0}$  converges to a limit  $p_i^* \in (0, 1)$ . So that for some  $k^* \in \mathbb{N}$ ,  $\forall k \geq k^*$ ,  $p_i(k) \in (0, 1)$ . As a consequence, equation (3) rewrites as

$$\frac{p_i(k+1) - p_i(k)}{p_i(k)(1-p_i(k))} + p_i(k) = r_i(k).$$

Taking the limit of the previous equation shows that  $(n_i^+(k)/n_i)_{k \geq 0}$  converges and is thus stationary and its limit satisfies  $\rho_i^* = p_i^*$ . Since  $n_i^+(k), n_i \in \mathbb{N}$  and  $n_i^+(k) \leq n_i$ , it is clear that  $p_i^* \in \mathcal{S}$ . ■

## B. Preservation of actions

In this subsection we investigate the conditions ensuring that the action  $q_i$  does not change over time. More precisely, we provide a criterion to detect when a group of agents sharing the same action will preserve it for all time. For the sake of simplicity we focus on  $q_i(0) = 0$  but using similar arguments the same results can be obtained for  $q_i(0) = 1$ .

*Lemma 5:* Let  $i \in \mathcal{V}$ , if  $n_i^-(k) \geq n_i^+(k)$  and  $q_i(k) = 0$  then  $q_i(k+1) = 0$ .

In the sequel we denote by  $|A|$  the cardinality of a set  $A$ .

*Definition 6:* We say a subset of agents  $A \subseteq \mathcal{V}$  is a *robust polarized cluster* if the following hold:

- $\forall i, j \in A, q_i(0) = q_j(0)$ ;
- $\forall i \in A, |N_i \cap A| \geq |N_i \setminus A|$ .

*Remark 3:* Notice that, in this section we do not make any assumption on the connectivity of the interaction graph. This means in particular that it may happen to have  $n_i = 0$  for some agents belonging to  $A$  above.

The next result explains why the word *robust* appears in the previous definition.

*Proposition 7:* If  $A$  is a robust polarized cluster with  $q_i(0) = 0$  for a certain  $i \in A$  then

- $\forall i \in A, \forall k \in \mathbb{N}, q_i(k) = 0$ ;
- $\forall i \in A, \lim_{k \rightarrow \infty} p_i(k) \leq \frac{1}{2}$ .

*Proof:* The proof will be done by induction. Let us remark first that following the first item of Definition 6 we have  $q_i(0) = 0, \forall i \in A$ . Let us assume that for a fixed  $k^*$  and  $\forall i \in A$  one has  $q_i(k^*) = 0$ . Let us also recall that the interaction graph under consideration is fixed. Therefore we still have that  $|N_i \cap A| \geq |N_i \setminus A|$ . Moreover,  $|N_i \cap A| \subseteq N_i^-(k^*)$  implying that  $n_i^-(k^*) \geq n_i^+(k^*), \forall i \in A$ . Thus, Lemma 5 yields that  $\forall i \in A$  one has  $q_i(k^*+1) = 0$  and the proof ends. ■

## C. Change and diffusion of actions

The goal of this subsection is twofold. First to provide conditions at a given time  $k \in \mathbb{N}$  guaranteeing that the action of a fixed agent  $i \in \mathcal{V}$  will change at time  $k+1$ . Secondly, we analyze the propagation/diffusion of the action of a robust polarized cluster inside the overall network. Due to the symmetry of reasonings we continue to focus only on one case  $q_i(0) = 0$  or  $q_i(0) = 1$ .

*Proposition 8:* Let  $i \in \mathcal{V}$  and  $k \in \mathbb{N}$  such that  $p_i(k) > \frac{1}{2}$  and  $n_i^-(k) > n_i^+(k)$ . Let  $\epsilon(n_i^+(k), n_i) \in \left(0, \frac{1}{2}\right)$  be the unique positive real solution of the equation:

$$x^3 - \left(\frac{1}{2} - r_i(k)\right)x^2 - \frac{3}{4}x + \left(\frac{1}{2} - r_i(k)\right)\frac{1}{4} = 0. \quad (13)$$

Then  $p_i(k+1) \leq \frac{1}{2}$  if and only if  $p_i(k) < \frac{1}{2} + \epsilon(n_i^+(k), n_i)$ .

The previous result states that an agent will change its action when it is influenced by more opposite actions, only if its

opinion is sufficiently close to the boundary between the actions. The notion of sufficiently closed depends on the proportion of opposite actions that influence the agent and is exactly quantified by  $\epsilon(n_i^+(k), n_i)$ .

*Lemma 9:* Let  $i \in \mathcal{V}$  and  $T_i \in \mathbb{N}$  such that  $n_i^-(k) > n_i^+(k), \forall k \geq T_i$ . Then it exists  $k^* \geq T_i$  such that  $q_i(k) = 0, \forall k \geq k^*$ .

The next result characterizes the diffusion of the action of a robust polarized cluster over the network.

*Proposition 10:* Let us consider the sets  $A_1, A_2, \dots, A_d$  such that

- $A_1$  is a robust polarized set with  $q_i(0) = 0$  for a certain  $i \in A_1$  (and thus  $\forall i \in A_1, q_i(0) = 0$ );
- $\forall h \in \{1, \dots, d-1\}$  and  $\forall i \in A_{h+1}$  one has

$$|N_i \cap A_h| > |N_i \setminus A_h|.$$

Then, for all  $h \in \{1, \dots, d\}, \exists T_h \in \mathbb{N}$ , such that,

$$\forall k \geq T_h, \forall i \in A_h, q_i(k) = 0,$$

where one can choose  $T_1 = 0$ . Consequently,  $\forall i \in \bigcup_{h=1}^d A_h$

one has  $\lim_{k \rightarrow \infty} p_i(k) \leq \frac{1}{2}$ .

*Proof:* The proposition with  $h = 1$  and  $T_1 = 0$  follows from the Proposition 7. For  $h > 1$ , we proceed recursively. Assume that the proposition holds for  $h \in \{1, \dots, f\}$  with  $f < d$  and show that it holds for  $h = f+1$ . We know that  $\forall i \in A_{f+1}$  one has

$$|N_i \cap A_f| > |N_i \setminus A_f|.$$

Moreover  $N_i \cap A_f \subseteq N_i^-(k), \forall k \geq T_f$  and  $\forall i \in A_{f+1}$ . Therefore we can apply Lemma 9 for any  $i \in A_{f+1}$ . Choosing  $T_{f+1} = \max_{i \in A_{f+1}} T_i$  one obtains that the proposition holds for  $h = f+1$ .

The last part of the statement is a simple consequence of

the fact that  $\forall i \in \bigcup_{h=1}^d A_h$  one has  $q_i(k) = 0, \forall k \geq T$  where  $T = \max_{h \in \{2, \dots, d\}} T_h$ . ■

## V. SOME PARTICULAR NETWORK TOPOLOGIES

### A. Complete graph

In this subsection we use previous results to completely characterize the opinion dynamics when the interactions are described by the complete graph.

*Proposition 11:* If  $n^-(0) > n^+(0)$  then  $\forall i \in \mathcal{V}$  the limit behavior of the opinion is given by  $\lim_{k \rightarrow \infty} p_i(k) = 0$ . Reversely,  $n^+(0) > n^-(0)$  implies  $\lim_{k \rightarrow \infty} p_i(k) = 1$ .

Let us consider now the case  $n^-(0) = n^+(0)$ . It is noteworthy that in this case  $n = |\mathcal{V}|$  is even and

- $\forall i \in N^+(0)$  one has  $|N_i \cap N^-(0)| = \frac{n}{2} > \frac{n}{2} - 1 = |N_i \setminus N^-(0)|$ .
- $\forall i \in N^-(0)$  one has  $|N_i \cap N^+(0)| = \frac{n}{2} > \frac{n}{2} - 1 = |N_i \setminus N^+(0)|$ .

If the initial conditions are not symmetric w.r.t.  $\frac{1}{2}$  an agent will cross from  $N^+$  to  $N^-$  (or reversely) and we recover the situation treated in Proposition 11. Therefore, to finish the analysis we give the following result that deals with initial conditions symmetrically displayed w.r.t.  $\frac{1}{2}$ . This case emphasizes an interesting oscillatory behavior of the opinions.

**Proposition 12:** Assume that  $n^+(0) = n^-(0)$  and moreover  $\forall i \in \{1, \dots, \frac{n}{2}\}$  there exist  $\eta_i(0) \in \left(0, \frac{1}{2}\right)$  such that

$$p_i(0) = \frac{1}{2} - \eta_i(0) \quad \text{and} \quad p_{\frac{n}{2}+i}(0) = \frac{1}{2} + \eta_i(0). \quad (14)$$

Then  $n^+(k) = n^-(k)$ ,  $\forall k \in \mathbb{N}$  and  $\exists k^* \in \mathbb{N}$  such that  $\forall k \geq k^*, \forall j \in \mathcal{V}$ ,

$$|p_j(k) - \frac{1}{2}| \leq \epsilon^* \quad \text{and} \quad \left(p_j(k) - \frac{1}{2}\right) \left(p_j(k+1) - \frac{1}{2}\right) < 0,$$

where  $\epsilon^* < \frac{1}{6(n-1)}$  is the unique positive solution of the equation

$$x^3 + \frac{1}{2(n-1)}x^2 + \frac{3}{4}x - \frac{1}{8(n-1)} = 0.$$

**Remark 4:** The result above states that all the agents in the network will finish by oscillating around  $\frac{1}{2}$  in a  $2\epsilon^*$  strip.

### B. Ring graph

Throughout this section we consider the particular configuration in which the interactions are described by an undirected graph in which each vertex has exactly two neighbors. In the following we identify agent  $n+1$  as agent 1 and agent 0 as agent  $n$ . For a precise representation of the graph we assume that  $\forall i \in \{1, \dots, n\}$  one has  $N_i = \{i-1, i+1\}$ .

**Proposition 13:** Under the ring graph topology the opinions dynamics (3) leads to the following properties:

- the set  $\mathcal{S}$  defined in (7) reduces to  $\left\{0, \frac{1}{2}, 1\right\}$ ;
- if  $\exists i \in \mathcal{V}$  such that  $q_i(0) = q_{i+1}(0) = 0$  then  $\{i, i+1\}$  is a robust polarized cluster (i.e.  $q_i(k) = q_{i+1}(k)$ ,  $\forall k \in \mathbb{N}$ );
- if  $\forall i \in \mathcal{V}$  one has  $q_i(0) = 1 - q_{i+1}(0)$  then
  - either the initial opinions are not symmetric w.r.t.  $\frac{1}{2}$  and agents will change actions asynchronously leading to robust polarized sets  $\{i-1, i, i+1\}$ .
  - or  $p_i(0) \in \left\{\frac{1}{2} - \sigma, \frac{1}{2} + \sigma\right\}$ ,  $\forall i \in \mathcal{V}$  and agents will change actions synchronously preserving  $n^-(k) = n^+(k)$ ,  $\forall k \in \mathbb{N}$ . Moreover, for  $\sigma$  solving

$$8\sigma^3 + 8\sigma^2 + 14\sigma - 1 = 0$$

one has  $p_i(k) \in \left\{\frac{1}{2} - \sigma, \frac{1}{2} + \sigma\right\}$ ,  $\forall i \in \mathcal{V}, \forall k \in \mathbb{N}$

## VI. NUMERICAL ILLUSTRATIONS

### A. Influential minority

We illustrate that a well connected minority can *convince* a majority of agents located in the periphery of the interaction network. Proposition 10 can be used to predict the phenomenon given the topology of the social network and the initial actions only. Figure 2 illustrates this fact. From Proposition 10, taking  $A_1 = \{1, 2, 3, 4\}$ ,  $A_2 = \{5, 6, 7, 8\}$ ,  $A_3 = \{9\}$  and  $A_4 = \{10\}$ , we predict that all agents will converge to a state with action  $\lim q_i = 0$ . We see in Figure 2-B that initially agents 9 and 3 tend to approach  $1/2$  since they have neighbors equally distributed over  $1/2$ . This behavior changes when agent 8 passes the  $1/2$  threshold. Moreover, the action diffusion propagates to agent 10 even though it originally had no neighbor with  $q_i(0) = 0$ . The decrease of agent 10 towards 0 only starts when agent 9 passes the  $1/2$  threshold.

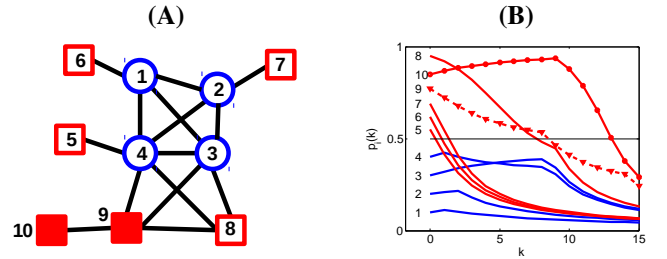


Fig. 2. Illustration of the action diffusion process described in Proposition 10. Agents 1, 2, 3 and 4 start with  $q_i(0) = 0$  and form a robust polarized cluster (Definition 6), while agents 5, 6, 7, 8, 9 and 10 start with  $q_i(0) = 1$ . All agents converge to a state with action  $\lim q_i = 0$ .

### B. CODA on a square lattice

We illustrate our results when the topology of interactions is a square lattice. First, we use a  $6 \times 6$  lattice (see Figure 3). As illustrated in Figure 3-B, for this type of structure, the smallest robust clusters are formed by  $2 \times 2$  squares. As expected from Proposition 7, the robust clusters keep their initial actions and patches of same actions form around the robust clusters. Notice also that the values of convergence lie in set  $\{0, 1/4, 1/3, 1/2, 2/3, 3/4, 1\}$  as predicted by Proposition 4.

Patches of agents with same actions are also observed for bigger lattice (see Figure 4 for an instance final actions in a  $50 \times 50$  lattice where initial conditions were drawn following independent uniform distributions.). This is in accordance with the patterns found in [8].

### C. Oscillatory dynamics on a ring graph

The following simulation displays the oscillation of agents' opinion around  $1/2$  when the interaction graph is complete and when the initial opinions are symmetrically distributed around  $1/2$  (see Proposition 12). Figure 5 shows an instance of this phenomenon for a system of 100 agents.

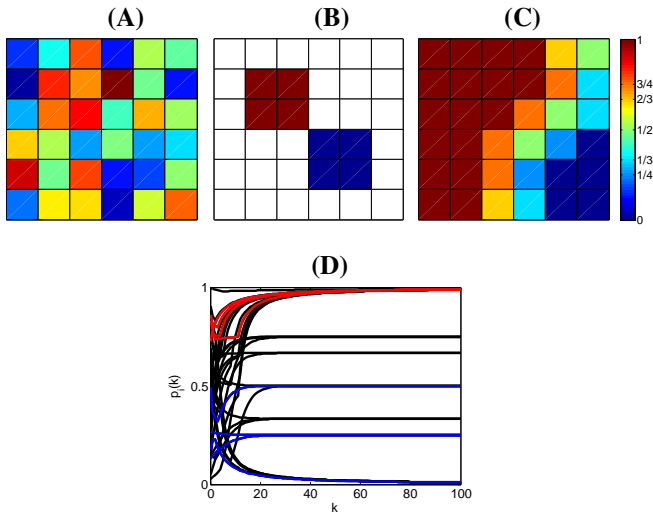


Fig. 3. Illustration of the CODA dynamics on a  $6 \times 6$  square lattice. (A) Each colored squared cell represents the initial opinion of an agent. Agents communicate with adjacent cells (above, below, left and right). (B) Robust clusters initially detected. (C) Final opinions of the agents. (D) Trajectories of the agents' opinions. Red lines correspond to agents belonging to a robust cluster with  $q_i(0) = 1$ . Blue lines correspond to agents belonging to a robust cluster with  $q_i(0) = 0$ .

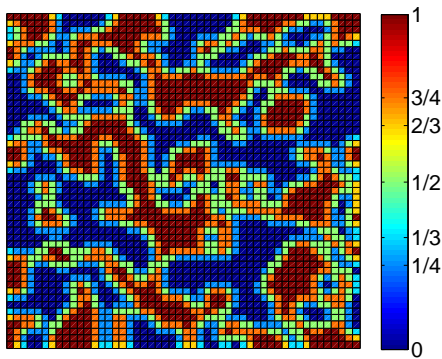


Fig. 4. Illustration of the CODA dynamics on a  $50 \times 50$  square lattice. Each colored squared cell represents the opinion of an agent after 100 iterations where initial conditions were drawn following independent uniform distributions. Agents communicate with adjacent cells (above, below, left and right).

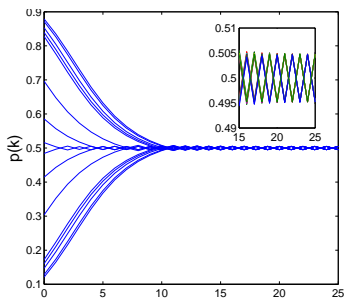


Fig. 5. Trajectories of 100 agents with the complete interaction graph and initial opinions distributed symmetrically around  $1/2$ .

## VII. CONCLUSIONS

In this paper we have introduced a novel opinion dynamics model in which agents have access to actions which are quantized version of the opinions of their neighbors. The

model reproduces different behaviors observed in social networks such as dissensus, clustering, oscillations, opinion propagation. The main results of the paper provides the characterization of preservation and diffusion of action under general communication topologies. A complete analysis of the opinions behavior is given in the particular cases of complete and ring communication graphs. Numerical examples illustrate the main features of this model.

## REFERENCES

- [1] C. Vande Kerckhove, S. Martin, P. Gend, P. J. Rentfrow, J. M. Hendrickx, and V. D. Blondel, "Modelling influence and opinion evolution in online collective behaviour," *PLoS ONE*, vol. 11, no. 6, pp. 1–25, 06 2016.
- [2] E. Ising, "Contribution to the theory of ferromagnetism," Ph.D. dissertation, University of Hamburg, 1924.
- [3] P. Clifford and A. Sudbury, "A model for spatial conflict," *Biometrika*, vol. 60, no. 3, pp. 581–588, 1973.
- [4] K. Sznajd-Weron and J. Sznajd, "Opinion evolution in closed community," *International Journal of Modern Physics C*, vol. 11, no. 6, pp. 1157–1165, 2000.
- [5] G. Deffuant, D. Neau, F. Amblard, and G. Weisbuch, "Mixing beliefs among interacting agents," *Advances in Complex Systems*, vol. 3, pp. 87–98, 2000.
- [6] R. Hegselmann and U. Krause, "Opinion dynamics and bounded confidence models, analysis, and simulation," *Journal of Artificial Societies and Social Simulation*, vol. 5, no. 3, 2002.
- [7] I.-C. Morărescu and A. Girard, "Opinion dynamics with decaying confidence: Application to community detection in graphs," *IEEE Transactions on Automatic Control*, vol. 56, no. 8, pp. 1862 – 1873, 2011.
- [8] A. Martins, "Continuous opinions and discrete actions in opinion dynamics problems," *International Journal of Modern Physics C*, vol. 19, no. 4, pp. 617–625, 2008.
- [9] F. Ceragioli and P. Frasca, "Continuous-time consensus dynamics with quantized all-to-all communication," in *Proceedings of European Control Conference*, 2015.
- [10] P. Frasca, "Continuous-time quantized consensus: Convergence of krasovskii solutions," *Systems & Control Letters*, vol. 61, no. 2, pp. 273–278, 2012.
- [11] S. Galam and S. Moscovici, "Towards a theory of collective phenomena: Consensus and attitude changes in groups," *European Journal of Social Psychology*, vol. 21, no. 1, pp. 49–74, 1991.
- [12] R. Axelrod, "The dissemination of culture: A model with local convergence and global polarization," *The Journal of Conflict Resolution*, vol. 41, no. 2, pp. 203–226, 1997.
- [13] S. Fortunato, V. Latora, A. Pluchino, and A. Rapisarda, "Vector opinion dynamics in a bounded confidence consensus model," *International Journal of Modern Physics C*, vol. 16, no. 10, pp. 1535–1551, 2005.
- [14] V. Blondel, J. M. Hendrickx, A. Olshevsky, J. Tsitsiklis *et al.*, "Convergence in multiagent coordination, consensus, and flocking," in *IEEE Conference on Decision and Control*, vol. 44, no. 3. IEEE; 1998, 2005, p. 2996.