# Metaphor of Politics: A Mechanism of Coalition Formation

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#### Abstract

Hybrid Multi-Agent Architectures support mobile robot colonies moving in dynamic, unpredictable and time varying environments to achieve collective team-oriented behaviors for solving complicate and difficult tasks. The development of a new coalition formation and coordination framework for robot colonies in dangerous, unknown and dynamic environment is outlined. The name of this new framework is Metaphor of Politics (MP), and it loosely takes inspiration from the political organizations of democratic governments. The main characteristic of the proposed framework lies in its dynamic reconfigurability in order to adapt the robot colony to environmental changes.

#### Introduction

The problem for the coalition formation and coordination of a robots team for complex tasks in dynamic, not predictable environments has been studied in the robotic literature by many researchers (Arkin and Balch, 1998) (Balch and Parker, 2002) (Gerkey and Mataric, 2002) (Fredslund and Mataric, 2002) (Mataric, 1995) (Murphy et al., 2002) (Parker, 1998). In this paper a new hybrid and dynamic framework is proposed. This framework takes inspiration from the political organizations of the democratic governments. The main idea behind the framework is that the leadership is not owned by a single robot, but by a government of robots. The "robot citizens" then execute the tasks according to the government rules. In this way a compromise may be reached among the centralized and the distributed approaches. The goal of the framework is to have a distribution of the planning actions, where each robot saves a deliberative independence status without losing its own reactivity. The agents receive high level goals by the government members and exploit their own deliberative capabilities to choose the faster strategy. The idea is to coordinate a colony of robots that are able to exhibit complex behaviors in order to accomplish a high-level mission but at the same time the framework includes a mechanism to form a new coalition

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caused by a failure of the government's strategy or a general inefficiency of the whole colony during the reaching of the mission's goals.

#### **Mathematical Model**

The MP framework considers a colony composed of H robots and M political parties, with M < H to guarantee the presence of at least one robot for every party. Within this framework a set of political issues is associated with every robot, which may express, for example, the individual's attitude towards risk, its dependence on reactive or deliberative behavior, its exploration proclivities, or interest in object recovery. The robots' attitudes towards these issues are represented over the range [-1,1], where 0 means don't care, -1 absolutely not, and 1 absolutely ves. Each party is represented by an ideal prototypical robot, standing for the central positions with respect to the political issues that characterize the party. Each robot is identified by Nfeatures; for each robot i and party j there is a vector of nissues:  $I_i^R, I_j^P \in M^{(n \times 1)}$  where  $i = 1 \dots H, j = 1 \dots M$ , P = party and R = robot. As an example to describe our model, we consider 3 issues which are identified with the following terms and meanings:

WELFARE: Energy of the robot
DEFENSE: Attitude towards risk
LABOR: Amount of work

Every issue is weighted by a non-negative coefficient (from 0 to  $+\infty$ ), where the coefficient represents the intensity or the strength of the issue. Every robot  $R_i$  and party  $P_j$  is represented by a vector with n components:

$$R_i = S_i^R \cdot I_i^R , \quad P_j = S_j^P \cdot I_j^P \tag{1}$$

where  $S_i^R$ ,  $S_j^P$  are diagonal  $n \times n$  matrixes containing the weights of the robots' attitudes towards the issues and of the parties issues respectively;  $R_i$  and  $P_j$  are representative of a robot and of a party in a multi-dimensional space called ROBOT ISSUES SPACE. The following example shows a specific situation of a colony made up 11 robots and 3 parties. Let us consider the following vectors of issues:

$$I_1 = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \qquad I_2 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \qquad I_3 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} , \qquad (2)$$

and the matrixes of weights:

$$W_{1} = \begin{bmatrix} 40 & 0 & 0 \\ 0 & 20 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad W_{2} = \begin{bmatrix} 20 & 0 & 0 \\ 0 & 20 & 0 \\ 0 & 0 & 20 \end{bmatrix}$$
(3)
$$W_{3} = \begin{bmatrix} 20 & 0 & 0 \\ 0 & 40 & 0 \\ 0 & 0 & 0 \end{bmatrix};$$
(4)

the political parties will be described by the following vectors:

$$P_{1} = \begin{bmatrix} 40 \\ -20 \\ 0 \end{bmatrix} \qquad P_{2} = \begin{bmatrix} 20 \\ 20 \\ 20 \end{bmatrix} \qquad P_{3} = \begin{bmatrix} -20 \\ 40 \\ 0 \end{bmatrix} . \tag{5}$$

The values of the components for each of the robots are set randomly in this case between 0 and 100:

$$R_{1} = \begin{bmatrix} 29.74 \\ -4.91 \\ 0.00 \end{bmatrix} \qquad R_{2} = \begin{bmatrix} 65.01 \\ -98.29 \\ 0.00 \end{bmatrix} \qquad R_{3} = \begin{bmatrix} 40.00 \\ -19.87 \\ 0.00 \end{bmatrix}$$
$$\begin{bmatrix} 73.33 \end{bmatrix} \qquad \begin{bmatrix} 41.98 \end{bmatrix} \qquad \begin{bmatrix} 91.99 \end{bmatrix}$$

$$R_4 = \begin{bmatrix} 73.33 \\ -37.58 \\ 0.00 \end{bmatrix} \qquad R_5 = \begin{bmatrix} 41.98 \\ 75.36 \\ 79.38 \end{bmatrix} \qquad R_6 = \begin{bmatrix} 91.99 \\ 84.47 \\ 36.77 \end{bmatrix}$$

$$R_7 = \begin{bmatrix} 62.08 \\ 73.12 \\ 19.38 \end{bmatrix} \qquad R_8 = \begin{bmatrix} -90.48 \\ 56.92 \\ 0.00 \end{bmatrix} \qquad R_9 = \begin{bmatrix} -23.44 \\ 54.87 \\ 0.00 \end{bmatrix}$$

$$R_{10} = \begin{bmatrix} -33.51 \\ 65.55 \\ 0.00 \end{bmatrix} \quad R_{11} = \begin{bmatrix} -62.73 \\ 69.90 \\ 0.00 \end{bmatrix}$$
(6)

Figure 1 shows a graphic representation of the robots and parties in the 3D space.

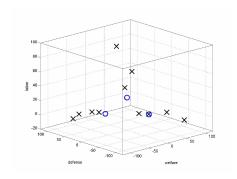


Figure 1: Graphic representation of robots and parties in Robot Issues Space

There exist several designated political roles within the coalition that a robot can occupy when elected. The heterogeneity of the robots inside the colony is described by a Roles Matrix (RM) which shows the capability for a robot to qualify for a role in the government; in the case of having H robots and 4 roles the matrix will be of rank  $H \times 4$ . For the example above, the roles matrix is as follows:

## **Voting Process**

The voting process can start for several reasons. It can be executed either periodically, in accordance with the elections of democratic governments, or in particular circumstances. An example of these anomalous situations is constituted by the failure of several robots, which would be a symptom of bad management by the government, and which would require new elections.

This process consists of two steps. The first step is *Cluster Identification*, where classical clustering techniques can be applied to our problem for the identification of the membership's groups. The literature provides several candidates; in particular we focus on *Voronoi tessellation*.

In the MP framework, the cluster identification step groups the robots of the colony on the basis of their party membership; this choice is based on the consideration that every robot maintains a political orientation depending on the closest aligned political party according to the issues. A robot  $R_i$  is deemed to belong to the  $P_j$  party if the following condition holds:

$$R_i \in P_j \Leftrightarrow d_{i,j} = \min_k \{d_{i,k}\} \quad k = 1, 2, \dots M \quad (7)$$

where  $d_{i,k}$  is the euclidean distance between the robot  $R_i$  and the idealized party position  $P_k$  in the ROBOT ISSUES SPACE (see figure 2). Table 1 shows the absolute distances

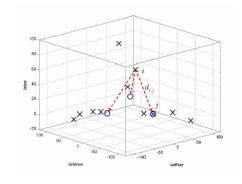


Figure 2: Distance calculation

between each robot and party, referring to the previous example, and the figure 3 shows the result of the clustering process.

	$\mathbf{P_1}$	$\mathbf{P_2}$	$P_3$
$\mathbf{R_1}$	18.24	33.40	67.02
$\mathbf{R_2}$	82.20	128.14	162.34
$R_3$	0.12	48.89	84.77
$\mathbf{R_4}$	37.69	81.00	121.37
$R_5$	124.10	84.12	106.75
$R_6$	122.35	98.09	126.00
$R_7$	97.65	67.78	90.61
$R_8$	151.47	118.20	72.48
$R_9$	21.57	59.19	15.27
$R_{10}$	112.80	73.07	28.91
$R_{11}$	136.52	98.67	52.16

Table 1: Absolute distances between robots and parties with respect to Issues

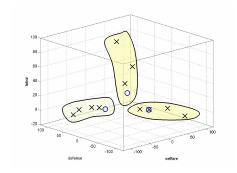


Figure 3: Clustering formation

The second step is *Vote Extraction*: the vote expressed by a robot is simulated by a random number generated in the interval [0,1]. This interval is divided into M sub-intervals, each associated to one of the M parties. The robot's generated value is considered as the vote expression for a particular party. For every robot i of the colony, the size of each party's subdivision of the interval [0,1] is based on the relative distances of the robot from each party j:

$$d_{i,j}^{REL} = \frac{\sum_{k \neq j} d_{i,k}}{\sum_{k} d_{i,k}} \% \qquad k = 1, 2, \dots M;$$
 (8)

To obtain  $\sum d_{i,j}^{REL}=100\%$  for each robot i, we normalize by a value equal to 1/(M-1). Table 2 shows these relative distances and figure 4 shows the division of the [0,1] interval for the robot  $R_1$ .

This representation allows the voting mechanism to employ *Monte Carlo* randomization methods. Table 3 shows the re-



Figure 4: Division of the [0,1] interval into 3 regions for the robot  $R_1$ 

(%)	$P_1$	$\mathbf{P_2}$	$P_3$
$\mathbf{R_1}$	42.31	35.92	21.76
$R_2$	38.97	32.80	28.22
$R_3$	49.95	31.72	18.31
$R_4$	42.14	33.12	24.72
$R_5$	30.29	36.64	33.05
$R_6$	32.34	35.84	31.81
$R_7$	30.93	36.76	32.30
$R_8$	27.86	32.72	39.40
$R_9$	21.57	32.85	45.57
$\mathbf{R_{10}}$	23.74	32.98	43.27
$R_{11}$	26.24	32.83	40.92

Table 2: Relative distances between robots and parties

sult of the resulting vote extraction expressed by each robot.

Robot	Vote	Party	
$R_1$	$P_2$	$P_1$	
$R_2$	$P_1$	$P_1$	
$R_3$	$P_3$	$P_1$	
$R_4$	$P_2$	$P_1$	
$R_5$	$P_2$	$P_2$	
$R_6$	$P_2$	$P_2$	
$R_7$	$P_3$	$P_2$	
$R_8$	$P_2$	$P_3$	
$R_9$	$P_1$	$P_3$	
$R_{10}$	$P_1$	$P_3$	
$R_{11}$	$P_1$	$P_3$	

Table 3: Vote extraction

#### **Coalition Formation**

The formation of a political coalition which constitutes the new government is made with the support of a linear space, the POLITICAL IDEOLOGY SPACE. This space represents all the robots and the parties belonging to the ROBOT IS-SUES SPACE. The mapping between these two spaces is performed by a mapping function  $f(\cdot)$  operating between the two representation spaces that groups robots exhibiting similar voting tendencies as evidenced during the voting process. The mapping is based on the following considerations: analogous to an actual parliament, the coalition axis is divided into 3 sections, each representing the ideologies of the left, center and right political viewpoints. The origin point of this linear space is centered to coincide with the pure political center, resulting in negative values being associated with parties aligned to the left and positive values to those towards the right. Two functions,  $f_R(\cdot)$  and  $f_L(\cdot)$ , are introduced in order to determine, when applied to a  $P_i$  party, those aspects which respectively characterize trends to the left or right. Each evaluated party will have an overall political trend determined as a compromise between its constituent right and left views on the issues. We represent the overall position of the party  $P_i$  with  $p_i$  in the coalition (ideology) space using the following heuristic equation:

$$p_j = f(P_j) = f_R(P_j) - f_L(P_j)$$
 (9)

According to the previous equation, a positive value identifies a rightist party while a negative one identifies a leftist one; values closer to zero identify right-center or left-center parties. The functions are based on suitable vectors of coefficients which weigh the members of the party. The coefficient values are related to the opinions associated with the issues of the parties in order to identify their ideology in the ROBOT ISSUES SPACE. For a linear f, the functional mapping is:

$$p_j = M_R^T \cdot P_j - M_L^T \cdot P_j = (M_R - M_L)^T \cdot P_j$$
 (10)

where  $M_R, M_L \in M^{(n \times 1)}$ .

We introduce a scaling factor for the positioning of the parties to avoid their being dispersed, while keeping unchanged the relative distances. Referring to the previous example, assigning to matrixes  $M_R$  e  $M_L$  the following values:

$$M_R = \begin{bmatrix} 1\\2\\0 \end{bmatrix} \quad M_L = \begin{bmatrix} 2\\1\\0 \end{bmatrix} , \tag{11}$$

The mapping of the individual robots is made recursively with respect to the mapping of the parties which have been previously voted for: an initial positioning is identified through the vote expressed, where a robot which voted for a  $P_j$  party will be located within a region around  $p_j$  in the space of the coalitions; subsequent refinement of the position allows placing the robot itself to the right or to the left of the voted party's position.

Referring to the previous example and considering a scale factor equal to  $\frac{1}{10}$ , the robot mapping yields the values shown in table 4.

Robot	$ r_i $	
$R_1$	-0.21	
$R_2$	-6.51	
$R_3$	5.48	
$R_4$	-0.50	
$R_5$	0.52	
$R_6$	0.60	
$R_7$	6.56	
$R_8$	-0.73	
$R_9$	-6.60	
$R_{10}$	-6.69	
$R_{11}$	-6.84	

Table 4: Robot mapping values

A political mass  $m_{i,j}$  is associated with each robot i and represents its weight within the voted party j; the calculation of this mass is based on the following function:

$$m_{i,j} = \frac{\sum_{k \neq i} d_{k,j}}{\sum_{k} d_{k,j}} \quad \text{if } i \text{ voted for } j, 0 \text{ otherwise} \quad (12)$$

where the index k includes all the robots of the colony which expressed a vote for the j party.

Table 5 shows the numeric values obtained by the calculation of the political mass for each robot, using the previous example.

$(m_{i,j})$	$P_1$	$P_2$	$P_3$
$R_1$	0.00	0.23	0.00
$\mathbf{R_2}$	0.27	0.00	0.00
$R_3$	0.00	0.00	0.52
$\mathbf{R_4}$	0.00	0.20	0.00
$R_5$	0.00	0.20	0.00
$R_6$	0.00	0.19	0.00
$\mathbf{R_7}$	0.00	0.00	0.48
$R_8$	0.00	0.18	0.00
$\mathbf{R_9}$	0.26	0.00	0.00
$\mathbf{R_{10}}$	0.24	0.00	0.00
$\mathbf{R_{11}}$	0.23	0.00	0.00

Table 5: Political mass of every robot

Every party represented in the POLITICAL IDEOLOGY SPACE is characterized by a mass center dependent on the robots which expressed a vote for the party, and the political masses associated with the robots themselves. This center of mass is obtained using an analogous concept from classical physics:

$$r_{CM}^{(j)} = \frac{\sum_{i} m_{i,j} \cdot r_{i}}{\sum_{i} m_{i,j}}$$
 (13)

where the index i describes all the robots of the colony which voted j. Table 6 shows the values of the mass centers for our example while figure 5 shows a graphic representation of this situation.

When the mapping process is finished, the coalition which will constitute the new government is formed by the winning party (the one which took the greatest quantity of votes), adding then adjacent parties until more than 50% of the total votes cast are reached. The concept of adjacent party is related to the distance between the center of mass of the winning party and the center of mass of the remaining ones. This coalition formation is represented in figure 5 for a case involving 3 parties and 11 robots.

Table 6: Mass center calculation

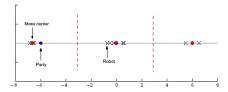


Figure 5: Mass center and coalitions

#### **Role Determination**

In our simulation, we choose the following government roles: Prime Minister (PM), Minister of Defence (MD), Minister of Communications (MC), assigned on the basis of the following rules: the PM is chosen from the robots belonging to the winning party, while the MD and the MC are chosen between the robots belonging to the winning coalition which have not assumed a previous governative role. Representing respectively with  $r_k^{(PM)}$ ,  $r_k^{(MD)}$  and  $r_k^{(MC)}$  the positions of the robots which satisfy these conditions, and with  $r_{CM}$  the position of the center of mass of the winning party, the PM role is assigned to the robot i closer to the  $r_{CM}$  center of mass, the MD role is assigned to the robot i positioned to the rightmost extremity of the coalition, and the MC role is assigned to the robot j positioned to the leftmost extremity of the coalition:

$$R_i = PM \Leftrightarrow r_i^{(PM)} = \min_k |r_k^{(PM)} - r_{CM}|$$
 (14)

$$R_i = MD \quad \Leftrightarrow \quad r_i^{(MD)} = \max_k \left( r_k^{(MD)} - r_{CM} \right)$$
(15)

$$R_j = MC \Leftrightarrow r_j^{(MC)} = \min_k \left( r_k^{(MC)} - r_{CM} \right)$$
 (16)

Table 7 shows the result of the role assignment process using our previous example; a graphic representation of the roles assignment appears in figure 6 and in figure 7. Notice that the robot nearest to the mass center is not chosen to cover the PM role because the Roles Matrix does not allow this configuration.

Table 7: Selected Government Robot Roles

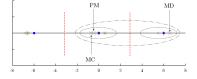


Figure 6: Determine roles in the POLITICAL IDEOLOGY SPACE

#### **Conducting Business**

The robots comprising the new government produce behavior to achieve their common goals in agreement with the

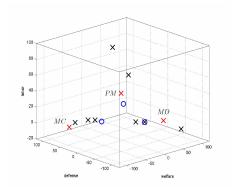


Figure 7: Determine roles in the ROBOT ISSUES SPACE

underlying political ideologies of the their coalition. The political ideologies are represented by a strategy that the robots must adopt. In the MP framework, two fundamental strategies are used: a leftist progressive (typically reactive) and a right-wing conservative (typically deliberative). In general, the government coalition is constituted by several parties for which these two strategies are the extremes of an overall methodology which changes its characteristics depending of the formation of the government. A rightcenter or left-center government will favor either a progressive or conservative strategy based on the weight given to the right or the left components during the formation of the government coalition. A strategy is characterized by a set of parameters which identify various aspects of the robot's behavior; each parameter has values along a continuous interval whose extremes (lower and upper) are associated with the left and right strategies. For every parameter s, for each of M parties there is an associated value so that  $s_1 \leq s_2 \leq \ldots \leq s_j \leq \ldots \leq s_M$  where j represents the generic party and  $s_1$ ,  $s_M$  identifies the two extreme parties. The winning coalition is constituted from M' parties with  $M' \leq M$ . The parameter  $s_c$  is only affected by the parties that form the coalition, where acts on the basis of its relative weight within in the coalition. Its value is calculated as a weighted average of the parameters of the coalition parties:

$$s_c = \sum_k a_k \cdot s_k \tag{17}$$

where k refers to the parties which form the coalition. The  $a_k$  weight associated with the k-th party is obtained by taking into account the  $V_k$  votes which it received with respect to the total votes of the coalition:

$$a_k = \frac{V_k}{\sum_h V_h} \tag{18}$$

Mini-Crisis. A mini-crisis is a mechanism which allows the partial replacement of the government with new robots belonging to the existing coalition and business is conducted using the same strategy. This mechanism eliminates the need for re-election of a new government and works to solve inefficiencies like the death, damage, or excessive loss of energy of any current government members, which would

negatively affect the behavior of the entire colony. A robot fault/failure requires a change in the Roles Matrix; for instance if the elected robot i-th cannot cover its role k-th any longer, then the matrix element (k,i) is replaced with a zero value. A mini-crisis is generated when an operating parameter exceeds its limits as well, for instance when the Welfare (which could provide information about the robots' available energy), falls below a critical threshold. A mini-crisis is not as critical as a full re-election that would stress the existing coalition, rejecting the current strategy being used, even if the overall colony behavior was acceptable.

**Re-election**. The re-election mechanism allows the colony to either reconfirm the previous coalition or change it completely. A re-election is normally caused by the expiration of a fixed time assigned for the government to complete the entire mission (TIME OUT), or in exceptional circumstances for evident deficiencies in the performance of the colony (NO CONFIDENCE), that is not imputable to a single robot but rather to the governing strategy. For instance if minicrises occur frequently then there is something wrong in the adopted policy and a re-election needs to be conducted.

# **Experimental Results**

Unlike robotic architectures which employ complex central management of a simple-robot colony, the MP framework focuses on coordination of robots which each possess a high degree of autonomy. Every robot is able to express a vote to elect a party, in order to form a political coalition and to identify a governing strategy that permits the accomplishment of mission goals. The MP framework has been implemented using the *MissionLab* simulation environment developed at the Georgia Institute of Technology, including the behavioral states for:

- 1) ELECTION
- 2) DETERMINING ROLES
- 3) CONDUCTING BUSINESS

The interaction among the states of the framework is shown in figure 8. The core of the implemented framework is the ELECTION state which represents a macro-state containing the two following sub-states:

- a) VOTING PROCESS
- b) COALITION FORMATION

which are responsible respectively for the voting mechanism for all robots and the formation of the political coalitions which constitute new governments. The VOTING PROCESS is constructed with a set of parameters that characterize each robot; these parameters change during the mission's execution relative for to the quality of assigned task execution. Each robot can be assigned a specific low-level task if the robot's role is as a citizen (e.g., exploration, bomb defusing) or a high-level task if the robot is a government member (planning, communication). The COALITION FORMATION

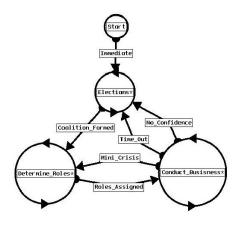


Figure 8: The Robot State Automata Diagram

phase determines the winning party on the basis of the votes expressed by the single robots. In order to constitute a winning coalition, parties are selected that are closely aligned to the winning one. The ELECTION state determines the winning political coalition which will constitute the new government. Once the elections have completed and the coalition has been formed, the DETERMINING ROLES state determines which robots will cover the governmental roles. Specifically PM, MD and MC are chosen in order to span the entire coalition ideology, thus guaranteeing the government's future strength. The CONDUCTING BUSINESS state allows a winning government to complete the various tasks assigned it, following a strategy which reflects the political trend of the coalition. The government's strategy is a compromise between the strategies of the single parties which compose the coalition. In the finite state automata (FSA) shown in figure 8, the triggers (links) produce the transitions between various states during mission execution. Their meaning is as follows:

- COALITION FORMED: Initiates robot role determination when election and coalition formation processes are concluded.
- ROLES ASSIGNED: Carries out the strategy embodied by the political trend of the government, occurring when the role assignment phase has finished.
- 3. MINI CRISIS: Makes a new assignment of government roles, while keeping the political trend of the government unchanged.
- 4. NO CONFIDENCE: Undertakes a new election as a result of general inefficiency in the whole robot colony.
- TIME OUT: Generates a new election process when the expiration of time assigned to the government to complete the whole mission occurs.

Figure 8 depicts the MP framework design for use in an unstructured, dynamic, and time-varying environment with

unknown or moving obstacles. This domain represents potential use in an application such as mine-defusing by a robotic colony with the presence of high risk for damage to the robots. In order to show MP performances, we created two alternative architectures, named Dictatorship and Anarchy. These alternatives had a twofold purpose: they are at the same time opponents and part of the MP framework. We used in fact them either individually to evaluate their own performance, or inside the MP architecture. In other words, we considered them the two extreme strategies (left and right) between whom our model can dynamically chose, as stated earlier.

## **Opponent Architectures**

These other approaches are distinguished from MP because of the total absence of elections, and thus the impossibility of changing their work in accordance with external conditions and previous results. They are now described in detail:

- 1. Anarchy: This architecture is characterized by the extreme uncertainty of its work. In fact, the robots belonging to the colony explore the area without any fixed strategy, until one of them finds a bomb. When it happens, the robot who found it, calls the nearest one to complete the requested operation. When this activity is completed (either by a successful defuse or with an explosion), all the robots restart their operations as before. Therefore, all the robots are equal in this approach, since each one pursues the same work, without any distinction in terms of roles.
- 2. Dictatorship: This second approach is instead characterized by a strong static distinction in terms of roles. In fact, the entire mission is coordinated by a single robot, called master, that is responsible for the choice of the supporter robot and the communication among the colony. It serves a crucial point for the mission, since its failure cannot be tolerated and would result in the failure of the entire mission. For this reason we tested this approach with three different values of fault probability for the master: 0%, 15% and 30%. Another important role is held by the supporters, the robots which are responsible for helping a robot who finds a mine. When it happens, the finder robot calls the master, who chooses the nearest supporter between the two supporters that are available. The last role is the searcher, that is a robot responsible for finding the mines scattered in the map.

## **Scenerio Descriptions**

The tests focused on a mission in which robots must discover mines that are scattered in the maps, and then they must defuse them. Each bomb has a probability to explode by accident, and a probability to activate its timer; so the defuse operation may not be successful. Moreover some enemy robots (terrorists), with each one being able to release a

certain number of mines, can also be present. In particular, four scenarios were developed:

- 1. 11 robots, no terrorists, with 30 mines placed in the map, time limit of 5 minutes. We took the following parameters: *Dead robots, Defused mines*
- 2. 11 robots, 2 terrorists that the robots can't kill, 5 mines placed in the map, time limit of 5 minutes. The following metrics were evaluated: *Dead robots, Defused mines*
- 3. 11 robots, 2 terrorists that the robots can kill, 5 mines placed in the map. The following metrics were used: *Dead robots, Mission Time, Time for Killing Enemies*
- 4. 11 robots, no terrorists, 25 mines placed in the map. The metrics include: *Dead robots, Mission Time*

#### **Environments**

To test the scenarios we used 3 different maps, intentionally developed; in particular they simulate a hangar (see figure 9), a market (see figure 10), and an airport (see figure 11).

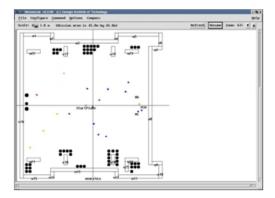


Figure 9: The Hangar Environment

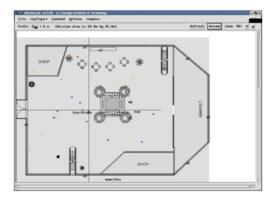


Figure 10: The Market Environment

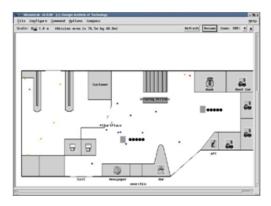


Figure 11: The Airport Environment

#### **Results**

In this section, several graphs are presented, the most significant ones, obtained by testing the various architectures in the different environments, and reporting either the results of each try, or the average achieved from a group of experiments made on a certain scenario and in a particular environment. The graphics illustrated in the next three figures refer respectively to the number of defused mines in the first scenario with a fault probability for Dictatorship's master of 0% (figure 12); the number of robots lost during the mission in the first scenario with a fault probability of 30% (figure 13); and the time spent for mission completion in the fourth scenery with a fault probability of 0% (figure 14), all in the airport map. In these graphs the numbers on the border indicate the number of the experimental group (where each group consists of one hundred experiments), while the numbers on the radii indicate the average value reported for each one. In all these figures it can be seen that the MP architec-

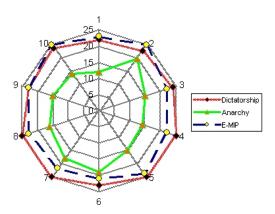


Figure 12: First Scenario: Bombs Defused

ture adapts its behaviors to the dynamic development of the mission, choosing from time to time the strategy that best adapts to external conditions, thanks above all to a coalition *regeneration* method. This allows the robots to confirm a

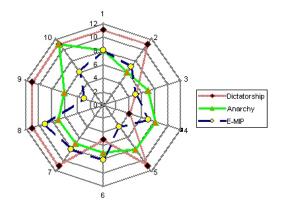


Figure 13: First Scenario: Dead Robots

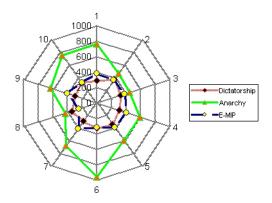


Figure 14: Fourth Scenario: Mission Time

governing coalition, if mission results are better when compared to previous governments. In fact it can be seen how the line representing MP performance always follows the best point between the opponent architectures, since it chooses the best mix according to their performances. In particular, in figure 12 the exterior line is the best, since it reports the highest values in terms of defused mines: in this case it is the anarchy strategy, but we can see the MP line is tightly close to it, since the architecture reflects what the best strategy is, and doesn't deviate from it often. In figure 13 the best line is the internal one, because it reports a lower number of dead robots. We see how opponent architectures alternate their performance, crossing themselves several times; MP is instead almost always in the middle, since it selecting among the best one at various times, thus maintaining good performance in all the experimental groups. In the end in figure 14, where the furthest internal line indicates lower mission time, thus better results, MP is very close to the Dictatorship line, because it obtains the highest performance during the experiments.

This phenomenon is even better depicted in the following two pictures, representing the average results, in terms of defused mines, obtained in the second scenario for the different cases of 0% Dictatorship's master fault probability (figure 15) and of 30% Dictatorship's master fault probability (figure 16) after the entire set of experiments. Here is even more evident how the MP architecture is able to choose the best strategy, because in the first case it selects Dictatorship, which has the best performance, but when the master's fault probability is increased thus reducing its result quality, MP robots quickly change the government coalition, moving their position towards the opposite side, thus maintaining enhanced performance, while the Dictatorship performance deteriorates.

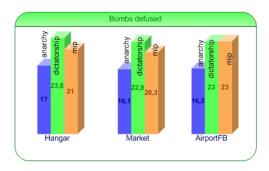


Figure 15: Second Scenario with Dictatorship's Master Fault Probability of 0%

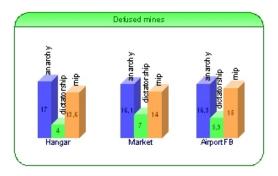


Figure 16: Second Scenario with Dictatorship's Master Fault Probability of 30%

## **Summary and Conclusions**

One of the fundamental and innovative features of the Metaphor of Politics Architecture is its dynamic social structure. It provides a good balance between the cost of forming a coalition and the performance of the coalition itself. It is also capable of forming superior political coalitions for difficult problem solving under conditions of limited information and resources.

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