Coordination in Human-Agent-Robot Teamwork

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ABSTRACT
Coordination is an essential ingredient of a teamwork-centered approach to autonomy. In this paper, we discuss some of the challenges and requirements for successful coordination, and briefly how we have used KAoS HART services framework to support coordination in a multi-team human-robot field exercise.

KEYWORDS: human-robot interaction, teamwork, coordination, KAoS, policies, OWL

1. TEAMWORK-CENTERED AUTONOMY
Planning technologies for intelligent systems often take an autonomy-centered approach, with representations, mechanisms, and algorithms that have been designed to accept a set of goals, and to generate and execute a complete plan in the most efficient and sound fashion possible. While this approach may be the best choice for situations where it is impractical or impossible for humans to provide close supervision of the intelligent system (e.g., [35]), it is not sufficient for the increasing number of applications that require close and continuous interaction with people and with other autonomous components (e.g., [5; 39]).

A teamwork-centered autonomy approach takes as a beginning premise that people are working in parallel alongside one or more autonomous systems, and hence adopts the stance that the processes of understanding, problem solving, and task execution are necessarily incremental, subject to negotiation, and forever tentative. Thus, a successful approach to teamwork-centered autonomy will require that autonomous systems be designed to facilitate the kind give-and-take and richness of interaction that characterize natural and effective teamwork among groups of people [8].

Over the past several years, we have been interested in learning how to facilitate such teamwork among humans, agents, and robots. To lay the groundwork for our research, we have studied how humans succeed and fail in joint activity requiring a high degree of interdependence among the participants [17; 31]. Such interdependence requires that, in addition to what team members do to accomplish the work itself, they also invest time and attention in making sure that distributed or sequenced tasks are appropriately coordinated.

Our research has been guided by three principles. First, we focus on situations where it is desirable for humans to remain “in-the-loop” and allow the degree and kind of control exercised by the human to vary at the initiative of the human or, optionally, with the help of adjustable autonomy mechanisms [10; 12; 30]. Second, we assure that mechanisms for appropriate robot regulation, communication, and feedback in such situations are included from the start in the foundations of system design, rather than layered on top as an afterthought [26]. Third, working in the tradition of previous agent teamwork researchers (e.g., [19; 41]), we attempt to implement a reusable model of teamwork involving a notion of shared knowledge, goals, and regulatory mechanisms that function as the glue that binds team members together. This teamwork model is to a large degree independent from and complementary to the set of domain-specific reasoners (e.g., task scheduling/optimization, spatial reasoning) that might be needed to accomplish a particular task objective.

Although there are several important challenges in making automation a team player [32], in this paper we focus on only the problem of coordination. Following a brief description of this aspect of joint activity, we describe the KAoS HART (Human-Agent-Robot Teamwork) services framework, which has been developed as a means of exploring our ideas about the role of regulatory constraints in joint activity [9; 11; 20; 26; 42; 43]. We give simple

1 There are important differences between human teams and the mixed teams of which we write, leading some to wonder whether the use of the term “team” is appropriate in this context and whether machines and software can appropriately be classed as “team members.” While recognizing the significant—and perhaps insurmountable—differences between the contributions that technology and people can make to joint activity, a large segment of the research community has concluded that using “team” is appropriate as a rough way of characterizing the ideal forms of interaction to which we aspire. For recent snapshots of this ongoing debate, see [3; 4; 15].
examples of some of the kinds of policies we have been exploring. Finally, we discuss a field exercise that allowed us to implement and explore many of these capabilities. This exercise involved mixed human-robot teams whose objective was to find and apprehend an intruder hiding on a cluttered Navy pier.

2. UNDERSTANDING COORDINATION

2.1. The Challenge of Human-Agent Coordination

Malone and Crowston [34] defined coordination as “managing dependencies between activities.” Teamwork, which by definition implies interdependence among the players, therefore requires some level of work for each party over and beyond the carrying out of task itself in order to manage its role in coordination. Part of that “extra” work involves each party doing its part to assure that relevant aspects of the agents and the situation are observable at an appropriate level of abstraction and using an effective style of interaction [6].

Although coordination is as much a requirement for agent-agent teamwork as it is for human-agent teamwork, the magnitude of the representational and reasoning gulf separating humans from agents is much larger. Moreover, because the agent’s ability to sense or infer information about the human environment and cognitive context is so limited, agent designers must find innovative ways to compensate for the fact that their agents are not situated in the human world. Brittness of agent capabilities is difficult to avoid because only certain aspects of the human environment and cognitive context can be represented in the agent, and the representation that is made cannot be “general purpose” but must include specific representations and optimizations for the particular use scenarios the designer originally envisioned. Without sufficient basis for shared situation awareness and mutual feedback, coordination among team members simply cannot take place, and, of course, this need for shared understanding and feedback increases as the size of the team and the degree of autonomy increase.

Notwithstanding these challenges, adult humans and radically less-abled entities (e.g., small children, dogs, video game characters) are capable of working together effectively in a variety of situations where a subjective experience of collaborative teaming is often maintained despite the magnitude of their differences. Generally this is due to the ability of humans to rapidly size up and adapt to the limitations of their teammates in relatively short order, an ability we would like to exploit in the design of approaches for human-agent teamwork.

2.2. The Elements of Effective Coordination

Basic requirements. There are three basic requirements for effective coordination: interpredictability, common ground, and directability [31]:

- Interpredictability: In highly interdependent activities, it becomes possible to plan one’s own actions (including coordination actions) only when what others will do can be accurately predicted. Skilled teams become interpredictable through shared knowledge and idiosyncratic coordination devices developed through extended experience in working together; bureaucracies with high turnover compensate for experience by substituting explicit, predesigned structured procedures and expectations.
- Common ground: Common ground refers to the pertinent mutual knowledge, beliefs, and assumptions that support interdependent actions in the context of a given joint activity [16]. This includes initial common ground prior to engaging in the joint activity, as well as mutual knowledge of shared history and current state that is obtained while the activity is underway. Unless I can make good assumptions about what you know and what you can do, we cannot effectively coordinate.
- Directability: Directability refers to the capacity for deliberately assessing and modifying the actions of the other parties in a joint activity as conditions and priorities change [14]. Effective coordination requires responsiveness of each participant to the influence of the others as the activity unfolds.

Order, predictability, and the origins of culture and social conventions. These three requirements are not just the basis for small group activity, but also the bedrock for human culture. Following the lead of pioneering researchers such as Geertz [23, pp. 44-46, 67], we have argued that people create cultures and social conventions—albeit in many disparate forms across mankind that can be hard for outsiders to understand—to provide order and predictability that lead to effective coordination [20; 21], including ongoing progress appraisal [22]. Order and predictability may have a basis in the simple cooperative act between two people, in which the parties “contract” to engage together in a set of interlinked, mutually beneficial activities. From this simple base, in humans at least, there are constructed elaborate and intricate systems of regulatory tools, from formal legal systems, to standards of professional practice, to norms of proper everyday behavior (along with associated methods of punishment or even simple forms of shaming for violations of these).

Coordination devices. People coordinate through signals and more complex messages of many sorts (e.g., face-to-face language, expressions, posture). Human signals are also mediated in many ways—for example, through third parties or through machines such as telephones or computers. Hence, direct and indirect party-to-party communication is one form of a coordination device, in

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1 Even simple forms of animal cooperation seems to bear out such a thesis [40], and we would argue that the more autonomous the agents involved, the more need there is for such regulation and the wider the variety of forms it might take.
this instance coordination by agreement. For example, a group of scientists working together on a grant proposal, may simply agree, through e-mail exchanges, to set up a subsequent conference call at a specific date and time. There are three other major types of coordination devices that people commonly employ: convention, precedent, and situational salience [17; 31].

Roles. Roles can be thought of as ways of packaging rights and obligations that go along with the necessary parts that participants play in joint activities. Knowing one’s own role and the roles of others in a joint activity establishes expectations about how others are likely to interact with us, and how we think we should interact with them. Shoppers expect cashiers to do certain things for them (e.g., total up the items and handle payment) and to treat them in a certain way (e.g., with cheerful courtesy), and cashiers have certain expectations of shoppers. When roles are well understood and regulatory devices are performing their proper function, observers are likely to describe the activity as highly-coordinated. On the other hand, violations of the expectations associated with roles and regulatory structures can result in confusion, frustration, anger, and a breakdown in coordination.

Organizations. Collections of roles are often grouped to form organizations. In addition to regulatory considerations at the level of individual roles, organizations themselves may also add their own rules, standards, traditions, and so forth, in order to establish a common culture that will smooth interaction among parties.

Knowing how roles undergird organizations and how rights and obligations are packaged into roles helps us understand how organizations can be seen as functional or dysfunctional. Whether hierarchical or heterarchical, fluid or relatively static, organizations are functional only to the extent that their associated regulatory devices and roles generally assist them in facilitating their constituent responsibilities and their work in coordinating their actions with others when necessary.

The lesson here for mixed human-agent-robot teams is that the various roles that team members assume in their work must include more than simple names for the role and algorithmic behavior to perform their individual tasks. They must also, to be successful, include regulatory structures that define the additional work of coordination associated with that role.

3. KAoS HART SERVICES FRAMEWORK

The KAoS HART (Human-Agent-Robot Teamwork) services framework has been adapted to provide the means for dynamic regulation on a variety of agent, robotic, Web services, Grid services, and traditional distributed computing platforms [9; 25; 26; 28; 33; 39; 42]. It also provides the basic services for distributed computing, including message transport and directory services, as well as more advanced features like domain and policy services.

3.1. KAoS Three-Level Architecture

Two important requirements for the KAoS architecture are modularity and extensibility. The result is a framework with well-defined interfaces that can be extended with components needed to support application-specific policies as required. The basic elements of the KAoS architecture are shown in Figure 1; its three layers of functionality, correspond to three different policy representations:

- **Human interface**: This is a hypertext-like graphical interface for policy specification in the form of natural English sentences. The vocabulary is automatically provided from ontology, consisting of highly-reusable core concepts, augmented by application-specific constructs in the ontology.

- **Policy Management representation**: This is used to encode and manage policy-related information in OWL (http://www.w3.org/TR/owl-features).

- **Policy Decision and Enforcement representation**: KAoS automatically “compiles” OWL policies to an efficient lookup format that provides the grounding of abstract ontology terms, connecting them to the instances in the runtime environment and to other policy-related information.

![Figure 1. KAoS Three-Layer Architecture](image)

3.2. Preliminary Studies

We began the process of developing a voice control ontology for robotic control by conducting an Oz study of a human operator controlling a mobile robot which has a camera for a search task. From this simple study we not only developed a very thorough ontology of voice commands that were sufficient for our final demonstration, but we also uncovered some interesting additional effects. First was the effect of the human’s perspective with the
robot [27]. We performed a follow on study to determine the impact of the operators interface on human-robot interaction. We also began a study on the difference between human-human interaction and human-robot interaction. The results covered all the typical specific teleoperation style commands such as “turn left 45 degrees” or “move forward five meters,” and more qualitative commands like “move forward slowly” or “speed up.”

3.3. Spatial Reasoning

People frequently use spatial references, especially when collocated and working on a task that is inherently spatial like a search task. We generated terminology from our Oz studies and then used this to produce an associated spatial reasoning ontology.

The KAoS Spatial Reasoning Component (KSPARC) is another component that augments the capabilities of team members. Operators can dynamically create objects based on sensor data and name them for future reference, for instance, by saying “this is a shed” and specifying the outline of an object on a GUI. Then, the operator can command a robot using object names as spatial references, for instance, “SILVER, move to the right side the shed.” When computing the destination position (i.e., a GPS point) from “the right side of …” robots that are not equipped with a sophisticated spatial reasoner of their own can rely on KSPARC. KSPARC can handle any perspective, because the interface uses a subject and an object. In this example case, the subject is the robot and the object is the shed. The robot could have taken the operator’s perspective by simply defining the subject to be the object instead. Multiple perspectives can be simultaneously accommodated (e.g., my right is your left). By providing such a capability as a service, we can deploy less capable robots as competent team members.

3.4. Dynamic Team Reasoning Mechanisms

All team members, human and agent, register with the KAoS directory service and provide a description of their capabilities. This enables team members to query the directory service to find specific team members, as well as match them based on capability. The domain and policy services manage the organizational structure among the agents, providing the specification of roles and allowing dynamic team formation and modification.

3.5. KAoS Robot Extension and Robot Behaviors

A “KAoS Robot” extension [26] provides a generic wrapper for each type of robot and a consistent interface for client systems to access the robots. KAoS Robot enables detailed status monitoring in addition to policy checking and enforcement, providing essential ingredients for coordination.

In order to adequately test new teamwork capabilities we wanted to employ, we needed to implement more complex and interesting behaviors. Hence, we enhanced our robot’s sensing capabilities with the addition of laser range finding and image processing for detection and tracking. We also increased the amount of datatypes that is observable to clients to include sonar, laser, images, mapping, and events such as finding and losing targets. We also added object detection and grouping, as well as mapping. Some of the more advanced behaviors included following, autonomous search of an area, and boundary monitoring. Lastly, we enabled the robots to make use of the KAoS HART services to provide their teamwork skills.

4. KAoS HART POLICY APPROACH

4.1. Teamwork Principles as Social Laws

The most popular approach among technologists has been to envision equipping such systems with social laws that would encourage “good” actions and discourage “bad” ones [9]. The idea of building strong social laws into intelligent systems can be traced at least back as far as the 1940s to the science fiction writings of Isaac Asimov [2].

In his well-known stories of the succeeding decades, he formulated a set of basic laws, designed to prevent harm to humans or other robots, that were built deeply into the “positronic-brain circuitry” of each robot so that it was physically prevented from transgressing them. Though the laws were simple and few, the stories attempted to demonstrate just how difficult they were to apply in various real-world situations. In most situations, although the robots usually were more consistent than humans in applying these “laws” and arguably behaved “logically,” they often failed to do the “right” thing, typically because the particular context of application required subtle adjustments of judgments on the part of the robot (e.g., determining which law took priority in a given situation, or what constituted helpful or harmful behavior).¹

Shoham and Tennenholtz [38] introduced the theme of social laws into the agent research community (see also [36; 44]), where investigations have continued under two main headings: norms and policies.

4.2. KAoS Policies

In contrast to Asimov’s laws of robotics—and similar in spirit to Grice’s famous maxims [24]—the objective of KAoS teamwork policies is not principally to prevent harm but rather, in a positive vein, to facilitate helpful interaction among teammates. As a body, we call these policies the “Golden Rules of HART,” recalling the biblical injunction: “Do unto others as you would have them do unto you” (Matthew 7:12). If robots (and people) were all sufficiently intelligent and benevolent, perhaps this abstract maxim would be the only rule needed for

¹ See [18] for an insightful essay concerning some of the implications of Asimov’s stories about the laws of robotics for information technologists.
coordination. Since this is not the case, a number of more specific instantiations of the general principle are required as the basis of teamwork policy. Through a relatively small number of such policies, we can help assure that complex emergent activity can remain coordinated. Through policy learning mechanisms, additional constraints may emerge in an adaptive manner [37].

KAoS policies expressing these specific coordination constraints are implemented in OWL (Web Ontology Language: http://www.w3.org/2004/OWL), to which we have added optional extensions to increase expressiveness (e.g., role-value maps) [42]. A growing set of services for policy deconfliction and analysis are also provided [7; 42].

Policies are used to dynamically regulate the behavior of system components without changing code or requiring the cooperation of the components being governed. By changing policies, a system can be continuously adjusted to accommodate variations in externally imposed constraints and environmental conditions. There are two main types of policies; authorizations and obligations. The set of permitted actions is determined by authorization policies that specify which actions an actor or set of actors are permitted (positive authorizations) or not allowed (negative authorizations) to perform in a given context. Obligation policies specify actions that an actor or set of actors is required to perform (positive obligations) or for which such a requirement is waived (negative obligations). From these primitive policy types, we build more complex structures that form the basis for team coordination.

4.3. Kaa and Kab: Adjustable Autonomy and Backup Plans

Rigid policy constraints and roles cannot cope with unanticipated situations in a dynamically changing environment. This is particularly important in teamwork situations where multiple agents have to cooperate to achieve a common goal. In addition to an agent’s own capabilities and constraints, we also need to take into account the capabilities and constraints of team members that might help or impede a joint task. IHMC’s Jung and Teng have devised a methodology to adjust a team of agents’ autonomy constraints and execution plans on the fly to avoid total task failures and performance degradation in teamwork situations [10; 12; 30].

The two components, Kaa (KAoS Adjustable Autonomy) and Kab (KAoS Adjustable Backups), are policy-based mechanisms that are built upon and integrated with KAoS services. The single agent model is extended to reason about coordination between multiple agents. Both qualitative and quantitative reasoning are employed to arrive at a satisfactory alternative plan in the situation when a task cannot be completed as planned (figure 2).

In the Coordinated Operation demo to be presented, Kaa and Kab provide adjustable autonomy services in unexpected situations such as device degradation and failure. For instance, when the camera fails in motion detection, Kab generates backup plans, replacing the camera with a less capable laser device or delegating the task with another robot equipped with a camera. Given the backup plans, Kaa performs decision-theoretic reasoning, computing the utilities of those backup plans. Based on the domain-specific knowledge, Kaa commands the robot to use its laser. Later, when the laser also fails, Kab generates another set of backup plans in the given situation, replacing the laser with a sonar or delegating the task to another robot. In this case, because the sonar’s performance in motion detection falls short of mission requirements, Kaa decides to delegate the current task (i.e., motion detection for securing boundaries) to another robot in order to maintain overall team performance. The field operation in which these maneuvers took place is taken up next.

5. THE COORDINATED OPS EXERCISE

5.1. Mission Scenario

Consider a scenario in which an intruder must be discovered and apprehended on a cluttered Navy pier (figure 3). To support the search, you can draw on the abilities of an additional human and five robots. While there are plenty of issues to address including robot capabilities, sensor limitations, and localization, we focused on the coordination aspects of the task. We specifically designed the task to have more robots than a single individual could easily handle by teleoperation. We also wanted to make sure the scenario included more than one human, since this provides its own challenges.1

Figure 2. Decision Mechanism of Kaa and Kab

5.2. The Exercise Environment

The pier involved the Agile Computing Infrastructure (ACI), the TRIPS dialogue-based collaborative problem solving system, Kaa/Kab adjustable autonomy and backup planning components, and an advanced multimodal display capability. For an overview of the entire system and scenario, see [29]

Figure 3. The pier

1 In addition to KAoS, which is our focus in this paper, the exercise involved the Agile Computing Infrastructure (ACI), the TRIPS dialogue-based collaborative problem solving system, Kaa/Kab adjustable autonomy and backup planning components, and an advanced multimodal display capability. For an overview of the entire system and scenario, see [29]
5.2. Team Composition

The available team members consisted of two humans and five robots (figure 4). The humans were to play distinct roles. One was the “Commander” who was to establish subteams and manage the overall search process. Relying on a combined speech and graphical interface, the Commander operated remotely without direct sight of the area of operation. The second human played the role of “Lieutenant.” The Lieutenant would be assigned to a team just like the robots, and he worked in the field generally alongside and in sight of them. He wore a backpack that carried a laptop to provide a similar speech and visual interface as the Commander’s, through a head mounted display as shown in figure 4. The robot team members included four Pioneer 3AT robots variously equipped with different combinations of sonar, GPS, pan-tilt-zoom cameras, and SICK lasers. The fifth robot was an IHMC-designed and -built robot called the tBot. All the robots had onboard computers and used wireless routers for communication.

![Figure 4. Initial two-tier hierarchical team structure](image)

All of the previously discussed policy sets, including acknowledgement, progress appraisal, notification, and chain of command, were in force for the exercise.

5.3. Mission Execution

The Commander first had to secure the area boundaries, and formed two subteams to block the two possible avenues of escape. Using natural language, the Commander composed two teams and assigned leaders for each of them (figure 5). One team (Team Alpha) was fully robotic, two robots with one assigned as the leader. The other team (Team Bravo) was mixed, two robots with the Lieutenant assigned to lead. Acknowledgement policies provided useful feedback to the Commander that teams had been successfully formed, since there was no external indication of the fact.

The Commander next defined an area of interest on his display and tasked each team to secure a particular side. After issuing the commands, the Commander dynamically created an obligation policy through speech to be notified by the team leaders when each team was in position. Once in position, the coordination policy took effect and the robot team leader reported.

![Figure 5. Three tier hierarchical team composed of two subteams (tBot still on original team).](image)

To apprehend the intruder, the Lieutenant tried to use the tBot, a robot not currently assigned to his team (figure 5). The coordination services enforced the chain of command and prevented the action. The Lieutenant then proceeded through the policy-required chain of command to acquire permission——i.e., he asked the Commander. The Commander dynamically assigned the tBot to the Lieutenant’s team. The Lieutenant was now authorized to make use of the tBot, and the apprehension was successful. Notice that the dynamic assignment of an agent to a certain group automatically brought with it all of that group’s extant regulatory structure, including the authority for that group’s leader to give orders to his new charge.

6. COORDINATION POLICY EXAMPLES

6.1. Modeling Teamwork for Coordinated Ops

The teamwork model for our coordinated operations exercise was implemented within various sets of KAoS policies. The intent of the policies is to provide information to establish and preserve common ground among both human and robotic team members, as well as helping to maintain organizational integrity. The policies are defined and enforced external to any specific robot API, so as new robots join, they automatically acquire all the teamwork intelligence possessed by the other robots. New capabilities supported by our coordinated operations policies included the following:
• Providing feedback by acknowledging commands, except when the relevant actions are directly observable.
• Only accepting commands from superiors in a chain of command. This helps prevent confusion and conflict among team members.
• Providing progress appraisal to the requestor of an action through notification when an action is finished, except when those actions are directly observable.
• Returning to base when the mission is complete or aborted and letting other teammates know
• Notifying the team leader when there is a status change.

During the demonstration, we showed how teams, roles, and policies can be dynamically-created, automatically deconflicted [7], and enforced at run-time.

6.2. Cohen-Levesque Notification Obligation Policy

One of the most well known heuristics in team coordination was originally formulated by Cohen and Levesque as follows: “any team member who discovers privately that a goal is impossible (has been achieved, or is irrelevant) should be left with a goal to make this fact known to the team as a whole” [19, p. 9]. We have implemented our version of this heuristic in the form of an obligation policy that can be roughly described as follows:

A Robot is obligated to notify its Teammates when Action is Finished (whether Successfully Completed, Aborted, or Irrecoverably Failed)

For example, in our field experiments, this policy ensured that, once an intruder has been apprehended, robot and human members of all teams, are notified [13]. This obligation would be triggered as soon as one robot became aware of this fact, and each robot would begin executing the appropriate task it was designed to perform following successful completion of the team goal (e.g., return to base, resume patrolling). If, on the other hand, the team commander were to abort the task due to a higher priority objective, or if any of the robots became aware that failure was inevitable, they would let their teammates know so that the appropriate behaviors for this situation would be triggered for the other members of the team. This single policy obviated the need to write a large number of special-purpose procedures for each possible success or failure mode.

6.3. Runtime Policy Addition and Modification

KAoS provides a mechanism to support runtime addition and modification of policies in support of coordination. For example, for a joint tracking task one partner may want to know when the other partner has acquired the target so he or she can disengage and reposition. Using the TRIPS dialogue capability [1] integrated with KAoS, the operator might simply state “Let me know when you see the target” in order to establish a one-time obligation. We have created a standing obligation policy for our robots that triggers a message stating “I see the target” when the target detection module determines that the target has been identified. Such a policy could be established by saying, “Always let me know when you see a target for the first time.”

6.4. Acknowledgements and Policy Deconfliction

We implemented a basic policy that requires robots to acknowledge requests. While this seemed a good general rule, there are important exceptions that need to be handled through KAoS policy deconfliction capabilities [7].

One reasonable exception to the acknowledgement policy is that people do not always verbally acknowledge requests, particularly when they are directly observable. Direct observability means that when a human requestor sends the communication to a robot receiver, the fact that the request was received, understood and being acted upon is observable by the requestor. For example, when a robot is told to move forward five meters, and then can be seen starting to move forward, there is normally no need for the robot to state “I have received your request to move forward and have begun.” The same applies to queries. When somebody asks a robot “where are you,” it is unnecessary for it to reply “I have heard your question and am about to reply”, if it, alternatively, simply says “in the library.” We implemented two additional policies to waive the obligation to acknowledge requests when the request is either a teleoperation command or a query.

Acknowledgement Policy Set
1) A Robot is obligated to acknowledge the Requestor when the Robot Accepts an Action
2) A Robot is not obligated to acknowledge Teleoperation requests
3) A Robot is not obligated to acknowledge Query requests

The two policies do indeed conflict with the original, but by assigning the more restrictive policies a higher priority (which can be done numerically or logically), it is possible to automatically deconflict these policies and achieve the desired behavior.

Note an additional advantage in the use of ontologies of behaviors is the fact that we can define policies for abstract classes of actions (e.g., Action, Teleoperation requests, Query requests) that will be enforced on every specific action that falls in that class.

6.5. Role Management and Progress Appraisal

Groups often use roles to perform task division and allocation. Roles provide a membership-based construct with which to associate sets of privileges (authorizations) and expected behaviors (obligations). When an actor is assigned to a role, the regulations associated with the role automatically apply to the actor and, likewise, are no longer applicable when the actor relinquishes the role. These privileges and expectations that comprise a role may
be highly domain dependent. For example the role “Team Leader” in a military domain is significantly different from “Team Leader” in sports. Roles may also specify expected behaviors. For example, if your role is a “Sentry,” then you are obligated to remain at your post, and other actors will expect you to fulfill that obligation. Roles can also affect other behaviors such as expected communications. If you are assigned to be a “Sentry,” you are obligated to announce any violations of your boundary and report these to your immediate superior.

Taking advantage of the extensibility and inheritance properties of OWL ontologies, we defined roles at various levels of abstraction with sub-roles refining the regulations pertinent to more generic super-roles. In this way, some high-level roles need not be domain specific or involve specific tasking, but they are still defined by their associated regulations. “Teammate” can be considered a generic role that has some of its regulations already noted. We view this level of abstraction as appropriate for expectations that facilitate coordination such as acknowledgements and progress appraisals. The obligation to acknowledge requests can be thought of as a policy associated with being a teammate. We have developed two policy sets that we feel apply generally to robots assigned to the role of “Teammate.” The first is the acknowledgement policy set discussed above. The second involves progress appraisal:

**Progress Appraisal Policy Set**

1) A Robot is obligated to notify the Requestor when requested Action is Finished (includes Completed, Aborted, and Failure).
2) A Robot is not obligated to notify the Requestor when a requested Tele-operation Action is Completed.
3) A Robot is not obligated to notify the Requestor when a requested Query Action is Completed.

The first policy ensures that the requestor of a task is notified when the tasked robot encounters problems or successfully completes the task since the action status of Finished is ontologically defined as a super-class of the statuses Completed, Failed, and Aborted. The second two policies in this set are exceptions similar to those in the acknowledgement set. With knowledge that these policies are in place, human and robotic team members have the mutual expectation that these progress appraisals will be performed. This interpredictability removes the need to explicitly ask for such communication and, perhaps just as importantly, the absence of these obligatory communications becomes an indicator that additional coordination may be necessary. For example, a robot is commanded to autonomously navigate to a distant location. Since it is known that the robot would notify team members if it had arrived, or it was stuck, or had otherwise failed, the others can assume that it is still moving toward the goal. If team members were concerned with an approaching deadline or that the task was taking too long, they would query for the robot’s position and create a new estimate of when it should reach the goal.

The policies outlined here are just two of several sets that we have explored, informed by previous theoretical work, simulations, and field experiments performed by ourselves and by others [1, 3, 15-17, 19, 26-28]. As we encounter new challenges in future work, we will continue to revise and expand such policy sets.

### 6.6. Policies Relating to Team Leaders

In contrast to our previous work on human-robot teams, where all team members were “equal,” we decided to explore the role of team “leaders.” Leaders not only must adhere to their own regulations, but they also impact the regulatory structure of all the other roles in the group. Peer interaction may be undirected, but Leaders tend to alter the pattern of activity, with themselves becoming the focal point. In particular we have identified several policy sets particular to leaders. The first set is about the chain of command:

**Chain of Command Policy Set**

1) A Robot is authorized to perform Actions requested by its Team Leader
2) A Robot is authorized to Accept Actions requested by a higher authority
3) A Robot is not authorized to perform Action requests from just any Requestor
4) A Robot is authorized to Accept Actions that are self-initiated

The first policy gives team leaders the authority to command their team. The second gives the same authority to anyone directly higher in the chain of command. The third policy explicitly restricts access to the robots from those outside of the chain of command. The fourth policy makes self initiated actions an exception to the third policy.

Another policy set was used to explore notification to help maintain common ground between the team leader and each of the team members:

**Notification Policy Set**

1) A Robot is obligated to notify its Team Leader when an Action is requested by a higher authority
2) A Robot is obligated to notify Its Team Leader when starting a self-initiated Action
3) A Robot is obligated to notify its Team Leader when a self-initiated Action is Finished (includes statuses of Completed, Aborted, and Failure).

### 6.7. Team Creation and Management

The KAoS Directory Service manages organizational structure, allowing dynamic team formation and modification. Teams and subteams can be created dynamically, allowing for the creation of complex organizational structures. Agents can join and leave teams as necessary to support the desired structure. Actors can be assigned roles including Team Leader, affecting the
dynamics of coordination as discussed in the previous section. Queries can be made to identify current team structure, who is on a certain team currently, or who is team leader.

7. CONCLUSIONS

The innovation in this research is not the hierarchical labeling of roles and team membership that normally goes as teamwork, but the regulatory infrastructure that affects behavior in a manner that facilitates teamwork. It is these regulatory obligations that define the team, much more so than a label or a colored jersey. In our work, the teams are not merely groupings, but provide the framework to support advanced coordination policies typical in human-human teams. When a leader is assigned, this means more then just being authorized to task other agents. For instance, it also defines the expected communication pattern among pertinent team members. As a team member, you are obligated to ensure that your leader knows you are working and to keep other members updated about pertinent information. These types of coordination, natural to humans, will enable robots to perform more like teammates are expected to act.

REFERENCES

34. Malone, T.W. and K. Crowston. "What is coordination theory and how can it help design cooperative work systems?" Presented at the Conference on Computer-Supported Cooperative Work (CSCW '90), Los Angeles, CA, 7-10 October, 1990, 357-370.