Goal Formation through Interaction in the Situation Calculus: A Formal Account Grounded in Behavioral Science

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ABSTRACT

Goal reasoning has been attracting much attention in AI recently. In this paper, we consider how an agent changes its goals as a result of interaction with humans and peers. In particular, we draw upon a model developed in Behavioral Science, the *Elementary Pragmatic Model (EPM)*. We show how the EPM principles can be incorporated into a sophisticated theory of goal change based on the Situation Calculus. The resulting logical theory supports agents with a wide variety of relational styles, including some that we may consider irrational or creative. This lays the foundations for building autonomous agents that interact with humans in a rich and realistic way, as required by advanced Human-AI collaboration applications.

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1 INTRODUCTION

Consider the following scenario: Barney is at home with Fred. Barney asks Fred what they should do. Fred answers by listing various options: play cards, watch a movie on a streaming service, have tea in the garden, or go out. Then Fred suggests to either play cards or watch a movie. Barney, on the other hand, proposes to either watch a movie or have tea in the garden. In the end they do neither of these things and actually go out.

If Barney and Fred were both human, then this would perhaps be a little unexpected, but not uncommon. But what if Fred was an artificial agent? If Fred was simply a virtual assistant, then this would indeed seem strange as such agents are normally subservient to their owner. But if Fred was instead a virtual companion, or a character in a game environment, then having such an option as a possible behavior would in fact be desirable, as it forms part of a rich believable human-like interaction. In this paper we examine this question, and develop a formal model of goal formation in interaction that can account for such behavior.

In recent years there has been growing interest in studying *goal reasoning*. In the words of [1], "intelligent systems may benefit from deliberating about, and changing their active goals when warranted. This flexibility may allow them to behave competently when they are not preencoded with a model that dictates what goals they should pursue in all encounterable situations." These ideas have lead to the

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so called *goal-driven architectures* [8, 37, 40], where goals have a central role in determining the behavior of an intelligent system. In such systems/agents, the need to revise goals typically arises because of external changes in the situation the agent is acting in. Such a change of circumstances forces the agent to revise her goals to make them rationally compatible, if possible, with the new situation [9].

Besides external changes, another common reason to change goals is interaction with a human or peer. The typical case studied is that of an agent getting an order from outside. In this case, the agent would typically *adopt the new goal as ordered*, possibly *dropping* her current goals (i.e., acting according to an *acceptor* relational style in EPM, see later), or alternatively adapting her previous goal to the new order (i.e., acting according to a *sharer* relational style in EPM). More recently the notion of *agent rebellion* has been considered, where the agent actually refuses to adopt the new goal [2] (i.e., acting according to a *maintainer* relational style in EPM).

However there are applications in which we want the intelligent system/agent to act more like an ordinary person, not an obedient soldier or even a soldier with her own ethics. Humans make use of a much wider set of relational styles to revise their goals as a result of an interaction, including ways that we may consider irrational, perhaps because they inflame conflicts, or exhibit creativity.

In this paper, we take this point seriously and examine a rich psychological model, called the *Elementary Pragmatic Model (EPM)* [14, 18]. EPM is inspired by the work of Bateson [5], and was developed in Psychiatry as a tool for family therapy. An important feature of EPM is that its principles are formulated mathematically. Leveraging EPM, we take a radical departure from previous work in the area and instead consider a rich set of possible *goal/desire formation mechanisms* as a direct result of *interaction with others*. Specifically, we show that we can take a recent advanced theory of goals dynamics [27, 29] and integrate into it the rich EPM set of relational styles as a set of interaction-based goal change mechanisms.

Apart from our technical proposal, this paper shows that current theories of goal dynamics are ready to accomodate rich goal formation mechanisms such as the one considered here. (We discuss other such theories in the final section.) This is quite important in view of having intelligent agents interacting with us in a more human-like manner, agents as companions rather than servants. Such capabilities are crucial in advanced Human-AI collaboration applications for personal welfare, including AI-based digital assistants, e.g., realistic chatbots [20], or interactive entertainment and believable agents [4, 24, 42], as well as social welfare, including counseling/coaching applications, and automated facilitators for group interaction [52]. Moreover, moving away from a naive view of goal formation/adoption is becoming more and more important with the development of autonomy in AI, in particular to avoid the

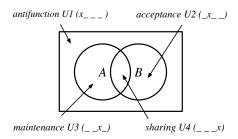


Figure 1: EPM coordinates

construction of Artificial Agents that act as fanatics even if mitigated by ethical principles, as recently advocated by Stuart Russell, who points out that agents should never be fanatically sure of the goals they are pursuing [45].

2 EPM AND RELATIONAL STYLES

In this paper we will follow a specific proposal: the Elementary Pragmatic Model (EPM), described mathematically in [48] and detailed with its clinical applications in [14, 18]. Like other models of the mind, EPM is a construction that promotes the study of psychological and psychopathological phenomena. EPM is rooted in the work of Gregory Bateson [5], which takes an interaction-based perspective on the human mind.

EPM is based on the idea that a subject's desires/goals change as a result of interaction: subject (i.e., her desire/goal) A changes to A' following interaction with the interlocutor (i.e., her proposal) B. The results of such an interaction is described in terms of four "coordinates", depicted as regions in the Venn diagram in Figure 1:

- *sharing* coordinate U_4 , $(__x)$, standing for $A \cap B$;
- maintenance coordinate U_3 , $(_x_)$, standing for $A \cap \bar{B}$;
- acceptance coordinate U_2 , ($_x_{_}$), standing for $\bar{A} \cap B$;
- antifunction coordinate U_1 , $(x_{_})$, standing for $\bar{A} \cap \bar{B}$.

Each coordinate may be set to 0 or 1, according to whether the related set is included or not in the result of the interaction. For example 0101 stands for $(\bar{A} \cap B) \cup (A \cap B)$, i.e., for B. The combination of such four coordinates gives rise to sixteen functions, $F0, F1, \ldots, F15$, which are called "relational styles". To illustrate how the sixteen functions work, consider the diagram in Figure 1. It has four regions, corresponding to the coordinates. The combinations deriving from systematically filling in one, two, three or four regions give rise to the sixteen possibilities, i.e., sixteen functions.

We illustrate the 16 relational styles with our "Barney and Fred" example from the introduction: Subject A (i.e., Fred) and interlocutor B (i.e., Barney) have the following options on what to do (for simplicity we assume these are the only options): "play cards", "watch a movie on a streaming service", "have tea in the garden", and "go out". Subject A expresses the desire to either "watch a movie" or "play cards". B instead suggests to A to either "watch a movie" or "have tea in the garden". We focus on A (i.e., Fred) and how his goals change through the interaction with B (i.e., Barney). The result of the interaction, based on each relational style is as described below:

F0 (0000) Void/Absent. *E.g.: Subject A decides they should not do anything.* No proposals, not even those which are shared with

the other will be accepted. The subject is unable to establish any relationship.

F1 (0001) Sharer. E.g.: Subject A decides they should "watch a movie". Only the proposals shared by both subjects are accepted.

F2 (0010) Acceptor of one's own world only. E.g.: Subject A decides they should "play cards". Only the proposals of the subject himself which are not shared with the other remain after the interaction has taken place. He refuses any overlapping element.

F3 (0011) Maintainer of one's own world. E.g.: Subject A keeps his own idea to either "watch a movie" or "play cards". All the elements of the subject remain unaltered by the interaction. Every element is maintained independently of the other's proposal.

F4 (0100) Acceptor of the other's world without sharing. *E.g.*: Subject A decides they should "have tea in the garden". The subject accepts proposals of the other only if they are not his own.

F5 (0101) Acceptor of the other's world. E.g.: Subject A changes his idea and is now willing to either "watch a movie" or "have tea in the garden". The subject substitutes for his own elements those of the other.

F6 (0110) Acceptor of one's own and of the other's world without sharing. *E.g.: Subject A is now willing to either "play cards" or "have tea in the garden"*. The subject keeps his own elements, accepts the other's, but drops the shared ones.

F7 (0111) Acceptor of one's own and of the other's world. E.g.: Subject A is now willing to either "watch a movie", "play cards" or "have tea in the garden". The subject is willing to accept the proposals of the other while maintaining his own.

F8 (1000) Acceptor of what only exists neither in one's own nor the other's world. E.g.: Subject A now considers undesirable all proposed activities "watch a movie", "play cards", and "have tea in the garden", and elects to "go out" instead. The subject loses his own elements, refuses proposals of the other and takes only elements outside of the ring of interaction, i.e., of the worlds of the two subjects, making a "creative" choice.²

F9 (1001) Acceptor of what only exists or does not exist, in one's own and in the other's world. E.g.: Subject A now considers undesirable "playing cards" and "having tea in the garden" and would like to either "go out" or "watch a movie". This is as in F8, except that the shared proposals by the interacting subjects are kept.

F10 (1010) Anti-other or "Mary-Mary quite contrary". E.g.: Subject A now considers undesirable "watching a movie" and "having tea in the garden" and would like to either "go out" or "play cards". The subject systematically refuses the other's proposals. In doing so, he even refuses elements from his own world.

F11 (1011) Complete maintainer of one's own world, with tendencies to expand. E.g.: Subject A maintains his interest in "watching a movie" and "playing cards", but now also considers "going out". In absolutely conserving its own point of view the subject also incorporates elements from outside of the ring of interaction.

F12 (1100) "Pseudoaltruist". E.g.: Subject A changes his mind and becomes interested in "having tea in the garden", but also in

 $^{^{1}}$ Notice that A still wants to do something with B (the context of the interaction) but none of the activities that A and B had originally in mind. In this case, the only remaining option is to "go out", but in general A could deliberate further to decide among the remaining choices.

²The relational styles F8–F15, which involve the antifunction coordinate, have been shown to be related to human creativity and have been utilized in tools for creativity development, see e.g.,[14, 15, 17, 46].

"going out". The subject rejects his own world, even the elements shared with the other, and accepts everything else. He could seem very complying with respect to the other, but he is "hiding an F3" through its complement.

F13 (1101) Exaggerated acceptor who refuses solely what exists in one's own world. E.g.: Subject A gets interested in "watching a movie" and "having tea in the garden", plus the never proposed option to "go out". He totally accepts the other's proposals as well as elements outside of the ring of interaction.

F14 (1110) Total acceptor who is nevertheless unable to share. E.g.: Subject A changes his mind and gets interested in "playing cards" and "having tea in the garden", but now considers interesting also "going out". The subject avoids sharing at the cost of accepting elements outside of his own world.³

F15 (1111) Total acceptor. *E.g.: Subject A loses any particular preference and now considers doing any activity.* The subject will accept any proposals even those not formulated. He says "yes" to everything. His behavior (like F0) doesn't produce any information.⁴

Below, we use EPM to develop a rich descriptive (vs. normative) formal model of goal change as a result of interaction.

3 FORMAL PRELIMINARIES

Situation Calculus. Our base framework for modeling goal change is the situation calculus [36] as formalized in [44]. In this framework, a possible state of the domain is represented by a situation. There is a set of initial situations corresponding to the ways the agents believe the domain might be initially, i.e., situations in which no actions have yet occurred. Init(s) means that s is an initial situation. The actual initial state is represented by a special constant S_0 . There is a distinguished binary function symbol do where do(a, s)denotes the successor situation to s resulting from performing the action a. Thus the situations can be viewed as a set of trees, where the root of each tree is an initial situation and the arcs represent actions. Relations (and functions) whose truth values vary from situation to situation, are called relational (functional, resp.) fluents, and are denoted by predicate (function, resp.) symbols taking a situation term as their last argument. There is a special predicate Poss(a, s)used to state that action a is executable in situation s.

We assume that we have a domain axiomatization that includes the following:⁵ (1) action precondition axioms, one per action a characterizing Poss(a, s), (2) successor state axioms, one per fluent, that succinctly encode both effect and frame axioms and specify exactly when the fluent changes [44], (3) initial state axioms describing what is true initially including the mental states of the agents, (4) unique name axioms for actions, and (5) domain-independent foundational axioms Σ describing the structure of situations [33].

Knowledge. Following [39, 47], we model knowledge using a possible worlds account adapted to the situation calculus. K(s', s)is used to denote that in situation s, the agent thinks that she could be in situation s'. Using K, the knowledge of an agent is defined as follows: $Know(\Phi, s) \stackrel{\text{def}}{=} \forall s'. K(s', s) \supset \Phi(s')$, i.e., the agent knows Φ in s if Φ holds in all of her K-accessible situations in s. Here and in the rest, Φ is a state formula, i.e., a formula with a single free situation variable, and $\Phi(s')$ is the formula that results from replacing free occurrences of this situation variable by s'. K is constrained to be reflexive, transitive, and Euclidean in the initial situation to capture the fact that agents' knowledge is true, and that agents have positive and negative introspection. As shown in [47], these constraints then continue to hold after any sequence of actions since they are preserved by the successor state axiom for K. We also assume that all actions are public, i.e., whenever an action (including exogenous events) occurs, the agent learns that it has happened. As in [27, 29], our framework supports knowledge expansion as a result of sensing actions [47] and some informing communicative actions.

Paths. Finally, to model general temporally extended properties (such as goals), we follow the approach of [27, 30], who extend the situation calculus with a new sort of paths, which are essentially infinite sequences of consecutive executable situations. We use variable p, possibly with annotations, to range over paths, and the special predicates OnPath(p, s) (resp. Starts(p, s)) to state that situation s is on path p (resp. is the first situation on path p). These are axiomatized as in [27, 30].

Golog. To represent and reason about complex actions or processes obtained by executing atomic actions, *high-level programming languages* have been defined. Here we concentrate on Golog [34], which includes the following constructs:

$$\delta ::= \alpha \mid \varphi? \mid \delta_1; \delta_2 \mid \delta_1 \mid \delta_2 \mid \pi x. \delta \mid \delta^*$$

In the above, α is an action term, possibly with parameters, and φ is situation-suppressed formula, that is, one with all situation arguments in fluents suppressed. As usual, we denote by $\varphi[s]$ the situation calculus formula obtained from φ by restoring the situation argument s into all fluents in φ . Program $\delta_1|\delta_2$ nondeterministically chooses between programs δ_1 and δ_2 . Program $\pi z.\delta(z)$ nondeterministically "picks" an object d to bind to variable z and then executes program $\delta(z)$ with z assigned to d. Program δ^* performs δ zero or more times. We can express if and while as (if φ then δ_1 else δ_2) \doteq $(\varphi?; \delta_1 \mid \neg \varphi?; \delta_2)$ and (while φ do δ) \doteq $((\varphi?; \delta)^*; \neg \varphi?)$, e.g., the program while $\exists x. \neg OnTable(x)$ do $\pi z. \neg OnTable(z)$?; table(z) repeatedly picks a block that is not on the table and tables it, until all blocks are on the table.

The semantics of Golog can be specified in terms of single-step transitions [13], using two predicates: (i) $Trans(\delta, s, \delta', s')$, which holds if one step of program δ in situation s may lead to situation s' with δ' remaining to be executed; and (ii) $Final(\delta, s)$, which holds if program δ may legally terminate in situation s. Using these predicates we can define whole computations as follows:

$$Do(\delta, s, s') \stackrel{\text{def}}{=} \exists \delta'. Trans^*(\delta, s, \delta', s') \land Final(\delta', s')$$

which says that by executing program δ in situation s we can get to situation s'. Also if we do not introduce concurrency, as here, we can define $Do(\delta, s, s')$ directly as in [34].

³Notice that this style of interaction has a *metaphoric flavor*: among the shared elements that are dropped one can consider all the concrete, everyday ones. What remains can describe concrete concepts only indirectly by drawing from either subjects' worlds or from outside them both (the presence of the antifunction coordinate adds an element of creativity) [14, 15, 17].

⁴F15 hints at a chaotic behavior, since everything is considered interesting; as Nietzsche said, "One must still have chaos in oneself to be able to give birth to a dancing star."

⁵We will be quantifying over formulae, and thus we also include axioms for encoding formulae and programs as first order terms, as in [13]; furthermore, we will also be using integers, and include an axiomatization of these as well.

4 A MODEL FOR GOAL DYNAMICS

Our model of goal change through interaction is based on Khan and Lespérance's (KL) situation calculus-based account of goals and their dynamics [27–29]. Note that KL handles temporally extended goals, since it incorporates a semantics based on infinite paths. Such semantics is analogous to the one for Linear-time Temporal Logic (LTL) [43], however the representation of the state is not propositional as in LTL, but fully first-order as in the Situation Calculus.

Prioritized Goals. KL formalize desires/goals with different priorities, which they call prioritized goals (p-goals, henceforth). These p-goals are not required to be mutually consistent and need not be actively pursued by the agent. Each p-goal has its own accessibility relation/fluent G. A path p is G-accessible (i.e., desirable) at priority level n in situation s, denoted by G(p, n, s), if all the goals of the agent at level n are satisfied over this path and if it starts with a situation that has the same history (in terms of the actions performed so far) as s. The latter requirement ensures that the agent's p-goal-accessible paths reflect the actions that have been performed so far. n ranges over natural numbers. A smaller n represents a higher priority, and the highest priority level is n0. Thus, it is assumed that the set of p-goals is totally ordered according to priority. One says that an agent has p-goal p0 at level p1 in situation p2 if and only if p3 holds over all paths that are p3-accessible at p3 in p5.

$$PGoal(\phi, n, s) \stackrel{\text{def}}{=} \forall p. G(p, n, s) \supset \phi(p).$$

Here and below, ϕ is a path formula, i.e., a formula with a single free path variable, and $\phi(p)$ is the formula that results from replacing free occurrences of this path variable by p.

Example. We can specify the initial p-goals of the agent in our running example as follows:

```
\begin{split} Init(s) \supset (G(p,0,s) &\equiv \exists s'.Starts(p,s') \land Init(s')) \land \\ (G(p,1,s) &\equiv \exists s'.Starts(p,s') \land Init(s') \land \exists s''.(s' \leq s'' \land OnPath(p,s'') \land \exists c.Activity(c) \land Doing(c,s''))) \land \\ (G(p,2,s) &\equiv \exists s'.Starts(p,s') \land Init(s') \land \exists s''.(s' \leq s'' \land OnPath(p,s'') \land \exists c.Activity(c) \land Doing(c,s'') \land \\ (c=WatchMovie \lor c=PlayCards))) \land \\ ((n>2) \supset (G(p,n,s) \equiv \exists s'.Starts(p,s') \land Init(s'))) \end{split}
```

That is, at the highest priority (level 0), the agent wants to be in an initial situation, then at the next highest priority (level 1), he wants to eventually be doing some activity (*Doing* is just an ordinary fluent), and then at the next highest priority (level 2), he wants to eventually be watching a movie or playing cards, and then at all lower priority levels (> 2), he wants to be in an initial situation. It follows that:

```
PGoal(\exists s'.Starts(p,s') \land Init(s')(p), 0, S_0) \land \\ PGoal(\exists s', s''.(Starts(p,s') \land s' \leq s'' \land OnPath(p,s'') \land \\ \exists c.Activity(c) \land Doing(c,s''))(p), 1, S_0) \land \\ PGoal(\exists s', s''.(Starts(p,s') \land s' \leq s'' \land OnPath(p,s'') \land \\ \exists c.Activity(c) \land Doing(c,s'') \land \\ (c=WatchMovie \lor c=PlayCards)))(p), 2, S_0) \land \\ (\forall n.n > 2 \supset PGoal(\exists s'.Starts(p,s') \land Init(s')(p), n, S_0))
```

Chosen Goals. In terms of the agent's p-goals, KL then define the agent's chosen goals or intentions (c-goals) that the agent is committed to and actually pursues. These are required to be consistent with each other and with the agent's knowledge, i.e., not ruled out by what is known. The agent's c-goals are essentially the largest set

of highest priority "realistic" p-goals that are consistent, where a given p-goal is preferred over all lower priority p-goals.

First, KL define *realistic* p-goal accessible paths:

$$G_R(p, n, s) \stackrel{\text{def}}{=} G(p, n, s) \wedge Starts(p, s') \wedge K(s', s),$$

i.e., paths that are G-accessible at n in s and start with a situation that is K-accessible in s. Thus G_R prunes out from G the paths that are known to be impossible.

Then, KL define the c-goal accessibility relation over paths $G_{\cap}(p, s)$, such that the agent has the c-goal that ϕ in situation s, i.e., $CGoal(\phi, s)$, if ϕ holds over all of her G_{\cap} -accessible paths in s:

$$CGoal(\phi, s) \stackrel{\text{def}}{=} \forall p. G_{\cap}(p, s) \supset \phi(p),$$

 $G_{\cap}(p,s)$ is in fact defined by induction on the priority level n, by first defining the paths that are in the maximal consistent set of highest priority "realistic" p-goals up to level n, $G_{\cap}(p,n,s)$, and then taking the c-goal accessible paths to be those for which $G_{\cap}(p,n,s)$ holds for all levels n:

$$G_{\cap}(p,s) \stackrel{\text{def}}{=} \forall n.G_{\cap}(p,n,s).$$

 $G_{\cap}(p, n, s)$ is axiomatized as follows:⁶

```
G_{\cap}(p, n, s) \equiv 
if n=0 then
if \exists p'. G_R(p', n, s) then G_R(p, n, s)
else Starts(p, s') \wedge K(s', s)
else if \exists p'. (G_R(p', n, s) \wedge G_{\cap}(p', n-1, s))
then (G_R(p, n, s) \wedge G_{\cap}(p, n-1, s))
else G_{\cap}(p, n-1, s).
```

That is, at level n = 0, $G_{\cap}(p, n, s)$ contains the G_R accessible paths at level 0 if there exist such a path (i.e., the agent's c-goals at level 0 are his RPGoals at level 0 if her PGoals at level 0 are consistent with what she knows), otherwise it contains all paths that start with a K-accessible situation (i.e., the agent's c-goals at level 0 are the trivial goal to be on a path where what he knows holds). At any level n > 0, $G_{\cap}(p, n, s)$ contains all the paths that are in G_{\cap} at the previous level n - 1 and are G_R accessible at level n if there exists such paths (i.e., the agent's c-goals at level n are her c-goals at level n - 1 plus her RPGoals at level n if the agent's PGoal at level n is consistent with what she know and her c-goals up to level n - 1), otherwise, it is simply the paths that are in G_{\cap} at the previous level n - 1 (i.e., the PGoal at level n is left out of the agent's c-goals because it is inconsistent with the agent's knowledge and higher priority goals).

Example (cont.) All the agent's initial p-goals are consistent, so he initially has the c-goal to (be in an initial situation and) eventually be watching a movie or playing cards:

```
CGoal(\exists s', s''.(Starts(p, s') \land Init(s') \land s' \leq s'' \land OnPath(p, s'') \land \exists c.Activity(c) \land Doing(c, s'') \land (c = WatchMovie \lor c = PlayCards)), S_0)
```

⁶if ϕ then ψ_1 else ψ_2 is an abbreviation for $(\phi \supset \psi_1) \land (\neg \phi \supset \psi_2)$.

⁷ Note that paths in a p-goal (i.e. desired), unlike those in a c-goal, need not to be consistent with that the agent knows, i.e., start with a *K*-accessible situation. They must however have the correct action history. Thus (as in KL), they only need to be realistic wrt the past action history, not the world state.

Subgoals. KL also account for the relationship between supergoals and subgoals. They take a p-goal ψ to be a *subgoal* of a p-goal ϕ in s iff ψ has lower priority than ϕ and ψ is also a p-goal at all levels where ϕ is a p-goal:

```
SubGoal(\psi, \phi, s) \stackrel{\text{def}}{=} \exists m.PGoal(\psi, n, s) \land \\ \exists n.PGoal(\phi, n, s) \land \neg PGoal(\psi, n, s) \land \\ \forall m.PGoal(\psi, m, s) \supset PGoal(\phi, m, s) \land m > n
```

In our example, the agent's p-goal to eventually be watching a movie or playing cards is a subgoal of that of eventually be doing some activity. As we will see below, this account guarantees several desirable properties of subgoal dynamics.

Basic Goal Dynamics. An agent's goals change when her knowledge changes as a result of the occurrence of an action (including exogenous events), or when she adopts or drops a goal. Here, we will mostly follow KL's formalization of basic goal dynamics, with one alteration: we will assume that every consistent (i.e., satisfied by some path) p-goal is a subgoal of another p-goal, where the "trivial" p-goal that the history of actions in the current situation has occurred is the root of the subgoal hierarchy (at priority 0). To ensure that this is the case, we require the initial state description to entail that $Init(s) \supset (G(p, 0, s) \equiv \exists s'.Starts(p, s') \land Init(s')))$, i.e., initially the root goal is the trivial goal to be in an initial situation. Note that the progression of this trivial goal will persist by the successor state axiom for G, see below. Given this assumption, it is sufficient to formalize two goal revision actions: $adopt(\psi, \phi, m)$, where the subgoal ψ is adopted relative to the parent goal ϕ at level m (which should be below the parent goal's level) and $drop(\phi)$, drop the goal ϕ . Note that p-goals that are primary are simply adopted relative to the trivial p-goal at the root of the hierarchy. The action precondition axioms for adopt and drop are as follows:

```
Poss(adopt(\psi, \phi, m), s) \equiv \exists n. PGoal(\phi, n, s) \land n < m, Poss(drop(\phi), s) \equiv True
```

That is, the agent can adopt the subgoal that ψ w.r.t. the parent goal ϕ at level m in s if she already has the p-goal that ϕ at priority greater than m in s, and can always drop the p-goal that ϕ . The dynamics of p-goals is specified through the *successor state axiom for G*:

```
\begin{split} G(p,\,n,\,do(a,\,s)) &\equiv \forall \phi,\,\psi,\,m.(a \neq a dopt(\psi,\,\phi,\,m) \land a \neq drop(\phi) \\ &\land Progressed(p,\,n,\,a,\,s)) \lor \\ \exists \phi,\,\psi.(a = a dopt(\psi,\,\phi,\,m) \land SubGoalAdopted(p,\,n,\,m,\,a,\,s,\,\psi,\,\phi) \\ \lor \exists \phi.(a = drop(\phi) \land Dropped(p,\,n,\,a,\,s,\,\phi)). \end{split}
```

Firstly, to handle the occurrence of a non-adopt/drop (i.e. regular) action a, one progresses all G-accessible paths to reflect the fact that this action has just happened; this is done using the Progressed(p, n, a, s) construct, which replaces each G-accessible path p' with starting situation s', by its suffix p provided that it starts with do(a, s'):

```
Progressed(p, n, a, s) \stackrel{\text{def}}{=} \exists p', s'. G(p', n, s) \land Starts(p', s') \land Suffix(p, p', do(a, s'))
Suffix(p, p', s) \stackrel{\text{def}}{=} OnPath(p', s) \land Starts(p, s) \land \forall s'. s \leq s' \supset (OnPath(p, s') \equiv OnPath(p', s'))
```

Any path over which the first action performed is not a is eliminated from the respective G accessibility level.

When adopting a subgoal, one must capture the dependencies between a goal and the subgoals and plans adopted to achieve it. In particular, subgoals and plans adopted to bring about a goal should be dropped when the parent goal becomes impossible, is achieved, or is dropped. KL handle this as follows: adopting a subgoal ψ w.r.t. a parent goal ϕ adds a new p-goal that contains both this subgoal and this parent goal, i.e., $\psi \wedge \phi$, at a priority lower than that of the parent, shifting down all the ones below. This ensures that when the parent goal is dropped, the subgoal is automatically dropped as well, since as we will see, when we drop the parent goal ϕ , we drop all the p-goals at all levels that imply ϕ including $\psi \wedge \phi$. Also, this means that dropping a subgoal does not necessarily drop the supergoal. This is formalized below:

```
SubGoalAdopted(p, n, m, a, s, \psi, \phi) \stackrel{\text{def}}{=}
\textbf{if } n < m \textbf{ then } Progressed(p, n, a, s)
\textbf{else } \textbf{if } n = m \textbf{ then}
\exists k.highestLevel(\phi, s) = k \land
Progressed(p, k, a, s) \land \psi(p)
\textbf{else } (n > m) \ Progressed(p, n - 1, a, s)
highestLevel(\phi, s) = k \stackrel{\text{def}}{=}
PGoal(\phi, k, s) \land \forall \ell . \ell < k \supset \neg PGoal(\phi, \ell, s)
```

 $highestLevel(\phi, s)$ finds the highest level k where ϕ is a PGoal.

To handle dropping a p-goal ϕ , one replaces the propositions that imply the dropped goal in the agent's goal hierarchy by the "trivial" proposition that the history of actions in the current situation has occurred. Thus in addition to progressing all G-accessible paths, one adds back all paths that have the same history as do(a,s) to the existing G-accessibility levels where the agent has the p-goal that ϕ :

```
Dropped(p, n, a, s, \phi) \stackrel{\text{def}}{=} 
 if PGoal(\phi, n, s) \text{ then} 
 \exists s'.Starts(p, s') \land SameHist(s', do(a, s)) 
 else Progressed(p, n, a, s).
```

Note also that procedural goal/subgoals can be handled by using Golog [34]: the goal to execute program δ now can be represented by the path formula $\exists s, s'.Starts(p, s) \land OnPath(p, s') \land Do(\delta, s, s')$.

Properties of Goal Dynamics. KL have shown several results about p-goal/c-goal dynamics. Let \mathcal{D}_{KL} be the set of axioms and definitions in the KL theory of "optimizing agents" [27], i.e., the foundational axioms of the situation calculus with knowledge, the axiomatization of paths, the axioms encoding formulas and programs as terms, and the axioms specifying goals and their dynamics as outlined above. Proposition 4.4.15 in [27] states that \mathcal{D}_{KL} entails that the agent no longer has the progression of a p-goal ϕ after dropping it, unless it is strongly inevitable: 10

```
\mathcal{D}_{KL} \models PGoal(\phi, n, s) \land \\ \neg StronglyInevitable(ProgOf(\phi, drop(\phi), p), do(drop(\phi), s)) \\ \supset \neg PGoal(ProgOf(\phi, drop(\phi), p), n, do(drop(\phi), s)) \\ \text{where } ProgOf(\phi, a, p) \stackrel{\text{def}}{=} \\ \exists p', s'.Starts(p', s') \land Suffix(p', p, do(a, s')) \land \phi(p')
```

⁸ In KL there are two adopt actions, $adopt(\phi, m)$ for adopting ϕ as a primary goal, i.e., not as a subgoal (where m is the level at which ϕ is adopted), and $adoptRelTo(\psi, \phi)$ for adopting a subgoal ψ w.r.t. a supergoal ϕ , (where ψ is always adopted at the level just below the parent goal ϕ). Our approach unifies these two forms of adoption.

 $^{^9\}mathrm{KL}$ assume that the subgoal is always adopted at the level immediately below that of the parent; we generalize this below.

 $^{^{10}}$ A condition ϕ is strongly inevitable in situation s iff ϕ holds for all paths that start in a situation with the same action history as s see [27] for the formal definition. Also $ProgOf(\phi, a, p)$ means that the progression of ϕ holds after action a over path p.

This result still holds here as we have not changed how the G relation is affected by drop. KL also show that dropping a supergoal results in all its subgoals being dropped as well, but this only applies in the downward direction, i.e., dropping a subgoal does not cause its supergoals to be dropped. This holds here as well.

Regarding the effects of *adopt*, we can show some new results for our modified goal dynamics axiomatization. Let \mathcal{D}_{KL}^+ be \mathcal{D}_{KL} with the successor state axiom for G and the precondition axiom for *adopt* replaced by the ones above. We can show that \mathcal{D}_{KL}^+ entails that the agent does have the p-goal that ψ at level m after adopting it as a subgoal of ϕ at that level (if executable):

THEOREM 4.1.

```
 \mathcal{D}^+_{KL} \models Poss(adopt(\psi, \phi, \textit{m}), s) \supset \\ PGoal(\psi, \textit{m}, do(adopt(\psi, \phi, \textit{m}), s))
```

PROOF (SKETCH). We can prove this in a similar way to Prop. 5.3.1 in [27], the main difference being that there the subgoal ψ is always adopted at the level immediately below that of the supergoal ϕ , while here it is adopted at level m. The antecedent $Poss(adopt(\psi, \phi, m), s)$ ensures that ϕ is a p-goal in s at a level higher than m (and $highestLevel(\phi, s)$ is well defined). The result follows by the successor state axiom for G and the definition of PGoal. \square

We can also show that the adopted subgoal is a c-goal provided it is consistent with higher priority c-goals (and its parent is a c-goal):

THEOREM 4.2.

```
 \begin{split} \mathcal{D}_{KL}^{+} &\models \exists n.PGoal(\phi, n, s) \land n < m \land \exists p.G(p, n, s) \land G_{\cap}(p, n, s) \land \\ & highestLevel(\phi, s) = n \land \neg CGoal(\exists p', s'.Starts(p, s') \land \\ & \textit{Suffix}(p', p, do(adopt(\psi, \phi, m), s')) \land \phi(p) \land \psi(p'), m - 1, s) \\ &\supset CGoal(\psi, m, do(adopt(\psi, \phi, m), s)) \end{split}
```

PROOF (SKETCH). We can prove this in a similar way to Prop. 5.3.2 in [27]. The idea is as follows. All p-goals above level m will be progressed when $adopt(\psi, \phi, m)$ occurs. Since ψ (after the adopt) is consistent with c-goals above level m in s, all the p-goals that are c-goals are also consistent with ψ (after the adopt). We can also show that p-goals that are not c-goals up to level m remain so after after the adopt. Thus the p-goal levels that are c-goals up to level m remain the same after the adopt. ψ is added as a p-goal at level m by the adopt. It follows that some G-accessible paths from level m will be included in G_{\cap} after the adopt and thus ψ is a c-goal at level m in $do(adopt(\psi, \phi, m), s)$.

KL also prove some results about the persistence of achievement p-goals and c-goals under certain conditions.

Note that actions that affect the agent's knowledge such as sensing and informing actions also lead to changes in c-goals, as these are intentions that must remain consistent with what is known. P-goals on the other hand are just desires, and need not be consistent with what is known; they are only progressed when an action occurs,

5 GOAL CHANGE THROUGH INTERACTION

Now let's formalize the changes in goals that occur as a result of an agent interacting with another depending on her relational style. We represent such an interaction as a complex action/program $interact(agt_2, \psi_1, \psi_2, F)$, where the subject with relational style F proposes her p-goal ψ_1 to the interlocutor agt_2 who instead proposes

his p-goal ψ_2 . We will require that ψ_1 and ψ_2 be proper proposals:

```
\begin{array}{l} ProperProposal(\psi) \stackrel{\text{def}}{=} \\ \forall .a_1 = drop(\phi_1) \land a_2 = adopt(\phi_2, \phi_3, k) \supset \\ (ProgOf(\psi, do(a_2, do(a_1, s)) \equiv ProgOf(\psi, do(a_1, s) \equiv \psi(s)) \end{array}
```

This essentially means that the proposal ψ does not talk about the immediate dynamics (next 2 steps) of the agent's goals.¹¹ This complex action is defined as follows:

```
interact(agt_2, \psi_1, \psi_2, F) \overset{\text{def}}{=} \\ \pi k.highestLevel(\psi_1) = k?; \\ applyAttitude(F, \psi_1, \psi_2, k) \\ applyAttitude(F, \psi_1, \psi_2, k) \overset{\text{def}}{=} \\ \textbf{if } F \in \{F_1, F_2, F_3\} \textbf{ then } drop(False) \textbf{ else } drop(\psi_1); \\ \pi \phi.mostSpecCompatSuperGoal(\psi_1, \psi_2) = \phi?; \\ adopt(map(F, \psi_1, \psi_2), \phi, k) \\ mostSpecCompatSuperGoal(\psi_1, \psi_2, s) = \phi \overset{\text{def}}{=} \\ \exists k. \forall p. G(p, k, s) \equiv \phi(p) \land \\ Subgoal(\psi_1, \phi, s) \land \neg PGoal(\neg \psi_2, k, s) \land \\ \forall k', \phi'. k < k' \land [\forall p. G(p, k', s) \equiv \phi'(p)] \supset \\ \neg (Subgoal(\psi_1, \phi', s) \land \neg PGoal(\neg \psi_2, k', s)) \\ map(F_j, \psi_1, \psi_2) \overset{\text{def}}{=} \\ (U_4(F_j) \land \psi_1 \land \psi_2) \lor (U_3(F_j) \land \psi_1 \land \neg \psi_2) \lor \\ (U_2(F_j) \land \neg \psi_1 \land \psi_2) \lor (U_1(F_j) \land \neg \psi_1 \land \neg \psi_2) \\ \end{cases}
```

Essentially, $interact(agt_2, \psi_1, \psi_2, F)$ amounts to executing $applyAttitude(F, \psi_1, \psi_2, k)$, where the level of the revised subgoal k is the highest level where ψ_1 is a p-goal of the agent. $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to the agent dropping the current subgoal ψ_1 (unless $F \in \{F_1, F_2, F_3\}$ and the agent maintains subgoal ψ_1 , in which case we do drop(False), which has no effects), and then adopting at level k a boolean function of ψ_1 and ψ_2 that depends on F, $map(F, \psi_1, \psi_2)$, relative to supergoal ϕ , the most specific supergoal of ψ_1 that is compatible with ψ_2 in the situation (the functions U_i simply project the i-th coordinate of the relational style F). The different cases of $applyAttitude(F, \psi_1, \psi_2, k)$ according to $map(F, \psi_1, \psi_2)$ capture how the relational style affects how goals change as a result of the interaction. It is easy to see that executing $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to executing the following program:

```
case(F) {
    0000
               drop(\psi_1); adopt(False, \phi, k)
               drop(False); adopt(\psi_1 \wedge \psi_2, \phi, k)
    0001
    0010
               drop(False); adopt(\psi_1 \land \neg \psi_2, \phi, k)
    0011
               drop(False); adopt(\psi_1, \phi, k)
    0100
               drop(\psi_1); adopt(\psi_2 \land \neg \psi_1, \phi, k)
    0101
               drop(\psi_1); adopt(\psi_2, \phi, k)
    0110
               drop(\psi_1); adopt((\psi_1 \lor \psi_2) \land \neg(\psi_1 \land \psi_2), \phi, k)
               drop(\psi_1); adopt(\psi_1 \lor \psi_2, \phi, k)
    0111
     1000
               drop(\psi_1); adopt(\neg \psi_1 \land \neg \psi_2, \phi, k)
     1001
               drop(\psi_1); adopt((\psi_1 \wedge \psi_2) \vee (\neg \psi_1 \wedge \neg \psi_2), \phi, k)
     1010
               drop(\psi_1); adopt(\neg \psi_2, \phi, k)
               drop(\psi_1); adopt(\psi_1 \vee \neg \psi_2), \phi, k)
     1011
     1100
               drop(\psi_1); adopt(\neg \psi_1, \phi, k)
    1101
               drop(\psi_1); adopt(\neg \psi_1 \lor \psi_2, \phi, k)
    1110
               drop(\psi_1); adopt(\neg \psi_1 \lor \neg \psi_2, \phi, k)
    1111
               drop(\psi_1); adopt(True, \phi, k) \}
```

¹¹This requirement is not essential, but it greatly simplifies the definition of *interact* and the statement of our theorems, as the proposals are not affected by progression over the goal change actions that implement *interact*.

Let's discuss these different cases of apply $Attitude(F, \psi_1, \psi_2, k)$.

First, let's consider the cases where the agent maintains and possibly refines her goal ψ_1 , i.e., F1, F2, and F3. When the agent's attitude is F1, she elects to share the interlocutor's goal, and thus we take $applyAttitude(F,\psi_1,\psi_2,k)$ to amount to $adopt(\psi_1 \wedge \psi_2,\phi,k)$, i.e., adopt the conjunction of the interlocutor's goal ψ_2 with the agent's original goal ψ_1 , at the same priority level as ψ_1 . The parent of the new goal is the most specific supergoal of ψ_1 that is compatible with ψ_2 . Note that we do drop(False), which has no effect, just for uniformity, so that in every case applyAttitude involves a drop followed by an adopt.

Example. $interact(agt_2, \psi_1, \psi_2, F1)$ with the subject proposing to eventually be watching a movie or playing cards and the interlocutor proposing to be watching a movie or having tea in the garden

```
\begin{array}{l} \psi_1 = \exists s''.(Starts(p,s') \land Init(s') \land s' \leq s'' \land \\ \exists c.Activity(c) \land Doing(c,s') \land \\ (c=WatchMovie \lor c=PlayCards))) \\ \psi_2 = \exists s''.(Starts(p,s') \land Init(s') \land s' \leq s'' \land \\ \exists c.Activity(c) \land Doing(c,s') \land \\ (c=WatchMovie \lor c=TeaInGarden))) \end{array}
```

amounts to performing apply $Attitude(F, \psi_1, \psi_2, k)$ with the most specific compatible supergoal

```
\phi = \exists s''.(Starts(p, s') \land Init(s') \land s' \leq s'' \land \exists c.Activity(c) \land Doing(c, s'))
```

and the level of the revised subgoal k=2, which amounts to $adopt(\psi_1 \wedge \psi_2, \phi, k)$, i.e., the "shared" p-goal to eventually be watching a movie at level 2 replacing her original p-goal ψ_1 .

When the agent's attitude is F2, she elects to maintain her goal while rejecting the interlocutor's goal, and thus we take $applyAttitude(F, \psi_1, \psi_2, k)$ to amount to $adopt(\phi_1 \land \neg \psi_2, \phi, k)$, i.e., adopt the conjunction of the negation of the interlocutor's goal ψ_2 with the agent's original goal ψ_1 , at the same priority level as ψ_1 , relative to the most specific compatible supergoal.

When the agent's attitude is F3, she elects to maintain her goal unchanged, without adopting the interlocutor's goal, and thus we take $applyAttitude(F, \psi_1, \psi_2, k)$ to amount to readopting her goal ψ_1 ; her goals remain essentially unchanged.

Secondly, let's consider the cases where the agent accepts the interlocutor's goal, i.e., F4, F5, F6, and F7. In these, unless ψ_1 implies ψ_2 the agent is relaxing her strict preference for ψ_1 , adopting instead the interlocutor's goal possibly in conjunction with her own goal ψ_1 . Thus for F5, where the agent accepts the interlocutor's goal without rejecting her own, $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to dropping ψ_1 and then adopting ψ_2 relative to the parent, at the same level as her original goal. For F7, the agent accepts the interlocutor's goal while maintaining hers, and $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to dropping ψ_1 and then adopting $\psi_1 \vee \psi_2$ relative to the parent, at the same level as her original goal. For F6, the agent accepts the interlocutor's goal while maintaining hers, but rejects both goals holding, and thus $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to dropping ψ_1 and then adopting $(\psi_1 \vee \psi_2) \wedge \neg(\psi_1 \wedge \psi_2)$ relative to the parent, at the same level as

her original goal. Finally, for F4, the agent accepts the interlocutor's goal while rejecting hers, and $applyAttitude(F, \psi_1, \psi_2, k)$ amounts to dropping ψ_1 and then adopting $\neg \psi_1 \wedge \psi_2$ relative to the parent, at the same level as her original goal.

Thirdly we have the so called anti-functions F8 to F15 where the agent is willing to accept paths that satisfy neither her original goal ψ_1 nor the interluctor's goal ψ_2 . This introduces a creative element, giving rise to goals that were unforeseen before the interaction. For instance for F8, the agent decides to pursue paths where neither her original goal ψ_1 nor the interluctor's goal ψ_2 , and apply $Attitude(F, \psi_1, \psi_2, k)$ amounts to dropping ψ_1 and then adopting $\neg \psi_1 \land \neg \psi_2$ relative to the parent, at the same level as her original goal. For our example, this amounts to adopting the goal of eventually going out (the domain theory includes unique names and domain closure axioms for the 4 activities). For F9, the agent decides to pursue paths where neither her original goal ψ_1 nor the interluctor's goal ψ_2 as well as paths where both goals hold, and applyAttitude (F, ψ_1, ψ_2, k) amounts to dropping ψ_1 and then adopting $(\neg \psi_1 \land \neg \psi_2) \lor (\psi_1 \land \psi_2)$ relative to the parent, at the same level as her original goal. Notice that in our formalization, although the original goal is dropped and replaced by some combination that includes $\neg \psi_1$ and $\neg \psi_2$, such a combination is still a subgoal of the most specific compatible supergoal of ψ_1 and ψ_2 . In other words the context of the interaction remains unchanged. This gives the context for the creative element, avoiding a disruptive reconsideration of unrelated goals. Note also that in many of theses cases, the subject has merely rulled out certain options (e.g., for F14, rulled out $\phi_1 \wedge \psi_2$) and must later decide how she wants to realize the parent goal ϕ under these constraints.

Note that case F0, i.e., 0000, and case F15, i.e., 1111, superficially look alike. In both, we are dopping the original goal remaining with the supergoal. However in F0 the level k simply disappears while in F15 the super goal is readopted at level k. Thus in F0 the agent does not have any goal at level k while in F15 it has the context as the goal at level k but without committing to any means to achieve it.

Let us show formally that *interact* has the right effects on the subject's p-goals in all of these cases. Let \mathcal{D}_{GFI} be \mathcal{D}_{KL}^+ augmented with the axiomatization of Golog in [13] and the axioms and definitions for goal formation through interaction presented in this section. First, we can show (using Theorem 4.1) that in all cases, after the interaction the agent has adopted the changed p-goal $map(F, \psi_1, \psi_2)$ obtained by applying the agent's relational style to the two proposals:

```
THEOREM 5.1. For any j=0,\ldots,15 \mathcal{D}_{GFI} \models highestLevel(\psi_1,s)=k \land \\ Do(interact(agt_2,\psi_1,\psi_2,F_j),s,do(a_2,do(a_1,s))) \\ \supset PGoal(map(F_j,\psi_1,\psi_2),k,do(a_2,do(a_1,s)))
```

We can also show that in all cases but F1, F2, and F3, the agent has dropped her proposed goal ψ_1 after the interaction unless it is strongly inevitable:

```
THEOREM 5.2. For any j = 0, 4, ..., 15
\mathcal{D}_{GFI} \models highestLevel(\psi_1, s) = k \land k' \neq k \land \\ \neg StronglyInevitable(\psi_1, do(a_1, s)) \land \\ Do(interact(agt_2, \psi_1, \psi_2, F_j), s, do(a_2, do(a_1, s))) \supset \\ \neg PGoal(\psi_1, k', do(a_2, do(a_1, s)))
```

 $^{^{12}}$ This essentially replaces the subject's original goal ψ_1 by $\psi_1 \wedge \psi_2$. A reasonable alternative would be to do $adopt(\psi_2,\psi_1,k+1)$, i.e., adopt the interlocutor's goal ψ_2 relative to the subject's own goal ψ_1 , at a priority level just below that of ψ_1 . The difference in this case would be that the subject's would retain a higher priority for ψ_1 compared to $\psi_1 \wedge \psi_2$ and might fall back to the former after renouncing the latter.

PROOF (SKETCH). We have that $Poss(drop(\psi_1), s)$ since $interact(agt_2, \psi_1, \psi_2, F_j)$ is executable in s. Then by Prop. 4.4.15 in [27] (discussed earlier in Sec. 4) and ψ_1 being a proper proposal, it follows that $\neg PGoal(\psi_1, k', do(a_1, s))$ for all k'. In the case where k' < k, by the successor state axiom for G, the G-accessible paths at level k' in $do(a_2, do(a_1, s))$ are simply progressions over a_2 of G-accessible paths in $do(a_1, s)$ at the same level. Thus $\neg PGoal(\psi_1, k', do(a_2, do(a_1, s)))$. The case for k' > k is similar, but the set of paths is shifted down one level.

This holds for all levels except that of the new goal replacing ψ_1 ; to show it for level k, we need additional conditions on ψ_2 . Also, we can show (by Theorem 4.2) that afterwards, $map(F, \psi_1, \psi_2)$ is a c-goal as well, if it is consistent with higher priority c-goals:

```
Theorem 5.3. For any j = 0, ..., 15
\mathcal{D}_{GFI} \models highestLevel(\psi_1, s) = k \land \\ Do(interact(agt_2, \psi_1, \psi_2, F_j), s, do(a_2, do(a_1, s))) \land \\ mostSpecCompatSuperGoal(\psi_1, \psi_2, do(a_1, s)) = \phi \land \\ \exists n.PGoal(\phi, n, do(a_1, s)) \land n < k \land \exists p.G(p, n, do(a_1, s)) \land \\ G_{\cap}(p, n, do(a_1, s)) \land highestLevel(\phi, s) = n \land \\ \neg CGoal(\neg \exists s_1, s_2, p'.Starts(p, do(a_1, s_1)) \land \\ Do(interact(agt_2, \psi_1, \psi_2, F_j), s_1, s_2) \land \phi(p) \land \\ Suffix(p', p, s_2) \land map(F, \psi_1, \psi_2)(p'), k - 1, do(a_1, s)) \\ \supset CGoal(map(F, \psi_1, \psi_2), k, do(a_2, do(a_1, s))).
```

6 DISCUSSION

In this paper, we have focussed on formalizing how an agent adopting a relational style for a certain interaction with an interlocutor changes her goals. But note that some experimental studies on humans have shown that they adopt all relational styles [14, 16]. Essentially these experiments show that humans adopt all relational styles according to a statistical "pattern" assigning a normalized weight to each of the functions $F0, F1, \ldots, F15$. Such a pattern tends to have a similar shape for all normal individuals with a predominance of F3 and F1 but with all functions with a non-zero weight. These results mean that perhaps our artificial agents should also interact using a distribution over all relational styles according to a similar pattern. However, while there are results on the distributions of relational styles for humans, no one has as yet done experiments on the frequency of switching from one relational style to the next. This is an important issue for building artificial agents that are believable and akin to humans in their interactions [4, 35, 42].

Another fundamental question is how do subjects' relational styles themselves evolve. In EPM, the relational styles $F0, F1, \ldots, F15$, adopted by the subject are themselves objects on which the sixteen functions can be applied. This gives a formal model within EPM of how relational styles change as a result of interactions. Thus we get a table (the *paradox table*) of 256 (16×16) possible changes that allows one to forecast of how interactions may alter the relational style of the interacting subjects. These have been used in clinical practice as a guide to the therapist on how to act towards the patient [14, 17–19, 32]. We can foresee the use of this dynamics of relational styles to improve computer mediated coaching and group facilitation applications, where a artificial agent guides the interaction to help resolve uncomfortable or conflict situations [19, 31].

Finally, we observe that some of the relational styles in EPM are linked to creativity. This has lead to research on the use of EPM to

develop creativity enhancing techniques [15, 17, 46]. So far this has been used for creativity enhancement in humans, but it could also be used to help develop artificial agents that display creativity.

There has been much work on various frameworks for representing agents' goals and their dynamics in recent years [12, 51]. Much of this work has been motivated by the need to support declarative goals in agent programming languages [11, 21–23], to ensure that plan execution is tied to the achievement of the associated goals; for instance, if a plan fails to achieve its goal, another plan can be selected, and if a goal is achieved serendipitously, the associated plans can be dropped. Most of these frameworks only handle restricted forms of temporally extended goals, and few provide a model-theoretic semantics. None provide notions like the EPM relational styles. The KL framework is very general, handles arbitrary temporally extended goals, and has a well developed semantics. Postulates for goal/intention revision in the presence of beliefs are proposed in [10, 26]. [27] discusses which of these hold in KL.

We used the KL framework as a foundation for our account of goal formation through interaction, but the essence of the account is not tied to this particular framework and should readily be adaptable to others. The main requirements are support for goal adoption and contraction, as well as a hierarchy of subgoals.

In conclusion, the technical contributions of this paper are as follows: first in Section 4, we have generalized of the subgoal adoption mechanism of KL to allow the priority of the new subgoal to be specified and proved some important properties of the resulting goal dynamics framework, and second in Section 5, we have formalized an account of goal change through interaction based on EPM in our framework and proved that it satisfies some key requirements. More generally, we have shown how one can incorporates the rich range of relational styles from EPM into formal accounts of goal changes, a contribution, which can be fruitful for the realization of new kinds of artificial agents that are not purely rational servants.

In future work, we would like to extend our account to model how an agent's relational style is selected depending on the situation and how it evolves. We would also like to refine our account to incorporate models of emotions [41, 49], trust [6, 50], and norms [3], and how they affect goal change in interactions. We also want to examine the relationship of our EPM-based approach with work on argumentation frameworks [25, 38] and communication protocols, although the EPM deals with more basic considerations, namely interaction styles and attitudes, which are not necessarily rational. Our notion of interaction, like the EPM one, is abstract. Clearly one major way of making interaction concrete is through dialog and using conceptual tools such as speech act theory. So one could integrate our framework with a dialog model and extract from the dialog the relational styles in the interaction. This is a very compelling avenue for further research and some work on EPM such as [7] provides a good starting point. Finally, we would like to evaluate the usefulness of the account in applications.

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