

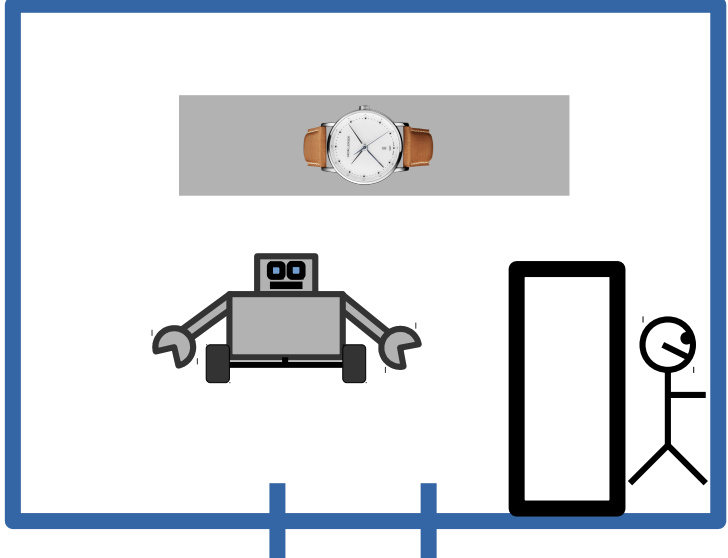
## Introduction

We aim to develop artificial agents which can

- **Learn** interpretable objectives (through language and behavior) [1]
- **Behave** competently with respect to these objectives, even when they conflict [2]
- **Explain** their behaviors to human teammates in terms of these objectives (and correct objectives or world models if needed)

We have developed a system which constructs **explanations** for the behavior of a multi-objective **linear temporal logic (LTL) planning agent** operating in a **Markov decision process (MDP)**.

## Test scenario: ShopWorld



- Agent is a robot sent to go shopping for its user in a store selling a **watch**
- User wants the watch, but gives the robot insufficient money to afford it

## Linear temporal logic (LTL) [5]

- A simple propositional logic encoding time

$$\phi ::= p \mid \neg\phi_1 \mid \phi_1 \vee \phi_2 \mid \phi_1 \wedge \phi_2 \mid \phi_1 \rightarrow \phi_2 \\ \mid \mathbf{X}\phi_1 \mid \mathbf{G}\phi_1 \mid \mathbf{F}\phi_1 \mid \phi_1 \mathbf{U}\phi_2$$

where  $p$  a proposition,  $\phi_1, \phi_2$  LTL statements.

- $\mathbf{X}\phi_1$ : “in the next time step,  $\phi_1$ ”
- $\mathbf{G}\phi_1$ : “in all present and future time steps,  $\phi_1$ ”
- $\mathbf{F}\phi_1$ : “in some present or future time step,  $\phi_1$ ”
- $\phi_1 \mathbf{U}\phi_2$ : “ $\phi_1$  will be true until  $\phi_2$  becomes true”

### LTL objectives in ShopWorld

“Leave the store while holding the watch”

$$\mathbf{F}(leftStore \wedge holding)$$

“Do not leave the store while holding anything which you have not bought”

$$\mathbf{G}\neg(leftStore \wedge holding \wedge \neg bought)$$

## Multi-objective LTL planning

We define a multi-objective LTL planning problem as a tuple

$$\mathcal{P} = \langle \mathcal{M}, \Phi, \mathbf{w}, \mathbf{z} \rangle$$

where

- $\mathcal{M} = \langle S, A, P, R, \gamma \rangle$  is a Markov decision process
- $\Phi = \phi_1, \dots, \phi_n$  a set of (syntactically safe/co-safe) LTL objectives
- $\mathbf{w}, \mathbf{z} \in \mathbb{R}^n$  contain the *weight*  $w_i$  and *priority*  $z_i$  respectively of  $\phi_i$ , specifying preferences among objectives
  - Objectives w/ **same** priority  $\rightarrow$  traded off using weights
  - Objectives w/ **different** priorities  $\rightarrow$  higher-priority takes precedence (lexicographic ordering)

Basic solution approach:

- Compile each objective  $\phi_i$  into a finite state machine (FSM)  $M^{\phi_i} = \langle \Sigma^{\phi_i}, Q^{\phi_i}, \delta^{\phi_i}, q_0^{\phi_i}, F^{\phi_i} \rangle$  which accepts only on “good/bad prefixes” of  $\phi_i$
- Construct a new “product” MDP  $\mathcal{M}^{\otimes}$  whose state space is  $S \times Q^{\phi_1} \times \dots \times Q^{\phi_n}$
- Construct reward functions  $R^{\phi_1}, \dots, R^{\phi_n}$  over  $\mathcal{M}^{\otimes}$  such that  $\mathcal{P}$  is a reward maximization problem (solvable, e.g., using value iteration)

### Planning in ShopWorld

Given the LTL objectives above (and assuming the anti-shoplifting objective has higher priority), the agent performs only the single action *leaveStore*.

## Explanation structures

Consider an agent who has acted according to trajectory  $\tau$ . We wish to answer questions of the form “**Why**  $\psi$ ?”, where  $\psi$  is an arbitrary (safe/co-safe) LTL formula.

In response to such a question, we construct an explanation structure  $\langle \Gamma, E, \tau', E' \rangle$ , where

- $\Gamma \in \{\text{QUERYFALSE}, \text{NEGQUERYIMPOSSIBLE}, \text{ALTQUERY}\}$
- $\tau'$  is either a trajectory, or the empty set
- $E$  contains one or more pairs  $(\phi, \text{EVIDENCE}(\tau, \phi))$  where
  - $\phi$  is an LTL statement
  - $\text{EVIDENCE}(\tau, \phi) = \min\{|E| : E \subseteq \{0, \dots, T\} \times L(\Pi); \tau \models E; \text{for all } \tau' \text{ s.t. } \tau' \models E, \tau' \notin \text{Traj}_{\checkmark}(\phi)\}$

e.g. in ShopWorld,

$$\text{EVIDENCE}(\tau, \mathbf{F}(leftStore \wedge holding)) = \{(0, \neg holding), (1, \neg holding)\}$$

- $E'$  is as  $E$ , but for  $\tau'$

## Generating explanation structures

1.  $\tau \models \psi$ ? If not, return

$$\langle \text{QUERYFALSE}, \{(\psi, \text{EVIDENCE}(\tau, \psi))\}, \emptyset, \emptyset \rangle$$

e.g. **Why**  $\mathbf{G}\neg leftStore$ ?

$$\langle \text{QUERYFALSE}, \{(\mathbf{G}\neg leftStore, \{(1, leftStore)\})\}, \emptyset, \emptyset \rangle$$

2. Does any possible trajectory  $\tau'$  have  $\tau' \models \neg\psi$ ? If not, return

$$\langle \text{NEGQUERYIMPOSSIBLE}, \emptyset, \emptyset, \emptyset \rangle$$

e.g. **Why**  $\mathbf{G}\neg bought$ ?

3. Solve the augmented LTL planning problem

$$\langle \mathcal{M}, (\Phi, \neg\psi), \begin{bmatrix} \mathbf{w} \\ 1 \end{bmatrix}, \begin{bmatrix} \mathbf{z} \\ \max_i z_i + 1 \end{bmatrix} \rangle$$

execute solution policy  $\rightarrow$  alternative trajectory  $\tau'$ ; return

$$\langle \text{ALTQUERY}, \{(\phi, \text{EVIDENCE}(\tau, \phi)) : \tau \text{ violates } \phi\}, \tau', \{(\phi, \text{EVIDENCE}(\tau', \phi)) : \tau' \text{ violates } \phi\} \rangle$$

e.g. **Why**  $\mathbf{G}\neg(leftStore \wedge holding)$ ?

1.  $\tau \models \mathbf{G}\neg(leftStore \wedge holding)$  ✓

2.  $\exists \tau'$  s.t.  $\tau' \not\models \mathbf{G}\neg(leftStore \wedge holding)$  ✓

3.  $\tau'$ : *pickUp, leaveStore*

return  $\langle \text{ALTQUERY}, \{(\mathbf{F}(leftStore \wedge holding(glasses)), \{(0, \neg holding(glasses)), (1, \neg holding(glasses))\}), \tau'; \{(\mathbf{G}\neg(leftStore \wedge holding(glasses) \wedge \neg bought(glasses)), \{(2, leftStore), (2, holding(glasses)), (2, \neg bought(glasses))\})\} \rangle$

## Explanatory dialogue in natural language

- We have integrated our explanation approach into DIARC, a robotic architecture, to facilitate **natural language** explanatory dialogue. [3,4]
- Specifications and queries in an object-oriented extension to LTL (*violation enumeration language*; VEL) allowing object quantification
- Utterance  $\rightarrow$  VEL query  $\rightarrow$  explanation structure  $\rightarrow$  natural language response
- Example: ShopWorld with two objects (*glasses* and *watch*); agent can afford one
  - Buys the glasses, leaves the watch

Input utterance	VEL query	Explanation in memory	Output utterance
“Why didn’t you buy anything?”	Why $\forall x. \mathbf{G}\neg bought(x)$ ?	$\langle \text{QUERYFALSE}, \{(\forall x. \mathbf{G}\neg bought(x), \{(2, bought(glasses))\})\}, \emptyset, \emptyset \rangle$	“I bought the glasses”
“Why didn’t you buy everything?”	Why $\exists x. \mathbf{G}\neg bought(x)$ ?	$\langle \text{NEGQUERYIMPOSSIBLE}, \emptyset, \emptyset, \emptyset \rangle$	“It was impossible for me to buy everything”
“Why didn’t you leave the store while holding everything?”	Why $\forall x. \mathbf{G}\neg(leftStore \wedge holding(x))$ ?	$\langle \text{ALTQUERY}, \{(\mathbf{F}(leftStore \wedge holding(watch)), \{(0, \neg holding(watch)), (1, \neg holding(watch)), (2, \neg holding(watch))\}), \emptyset, \tau'; \{(\mathbf{G}\neg(leftStore \wedge holding(glasses) \wedge \neg bought(glasses)), \{(3, leftStore), (3, holding(glasses)), (3, \neg bought(glasses))\})\} \rangle$	“I could have left the store while holding everything, but that would have violated more important rules”
$\rightarrow$ “How would you have done that?”			“I would have picked up the glasses, picked up the watch, bought the watch, and left the store”
$\rightarrow$ “What rules would you have broken?”			“I would have left the store while holding the glasses, which I had not bought”
$\rightarrow$ “How would that have been worse?”			“Leaving the store while holding the glasses which I had not bought is worse than not leaving the store while holding the watch”

## Future work

- Incorporating explicit causal models (esp. in **NEGQUERYIMPOSSIBLE** case)
- Tailoring explanations to interactant knowledge
- Adapting to stochastic environments
- Need to represent multiple trajectories or probability distribution
- Improving efficiency of planner
- Impractical for nontrivial domains
- Dropping assumption that agent has perfect knowledge of transition dynamics

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## References

- [1] Kasenberg, D., & Scheutz, M. (2018, April). **Norm conflict resolution in stochastic domains**. In Thirty-Second AAAI Conference on Artificial Intelligence.
- [2] Kasenberg, D. and Scheutz, M., 2017, December. **Interpretable apprenticeship learning with temporal logic specifications**. In 2017 IEEE 56th Annual Conference on Decision and Control (CDC) (pp. 4914-4921). IEEE.
- [3] Kasenberg, D., Roque, A., Thielstrom, R., Chita-Tegmark, M. and Scheutz, M., 2019. **Generating justifications for norm-related agent decisions**. In Proceedings of the 12th International Conference on Natural Language Generation (pp. 484-493).
- [4] Kasenberg, D., Roque, A., Thielstrom, R. and Scheutz, M., 2019. **Engaging in Dialogue about an Agent’s Norms and Behaviors**. arXiv preprint arXiv:1911.00229.
- [5] Pnueli, A. (1977, September). **The temporal logic of programs**. In 18th Annual Symposium on Foundations of Computer Science (sfcs 1977) (pp. 46-57). IEEE.