Uncertain Machine Ethics Planning

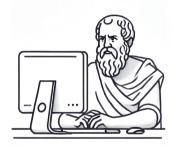
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A goal of Machine Ethics is to integrate ethical behaviour into autonomous decision making [Allen et al., 2006].

Adapting Philosophy of Ethics can be a challenge.

- Outcome uncertainty
- Moral uncertainty
- Expressive/practical for stakeholders
- Explainable/transparent



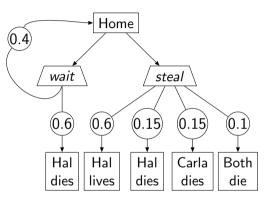
In [Kolker et al., 2023], we proposed Machine Ethics Hypothetical Retrospection...

- Based off Sven-Ove Hansson's Hypothetical Retrospection for ethical decision making under outcome uncertainty [Hansson, 2013].
- It is helpful to imagine our retrospection from major foreseeable outcomes, given information from decision time.
- MEHR systematises this for Machine Ethics with a simple argumentation procedure.

In this paper, we adapt and formalise MEHR for probabilistic planning.

Lost Insulin Running Example

- Hal is a diabetic who, through no fault of his own, has lost his insulin supply.
- He needs some urgently to stay alive.
- His neighbour, Carla, has some, but Hal does not have permission to take it.
- Is Hal justified in stealing to save his life?

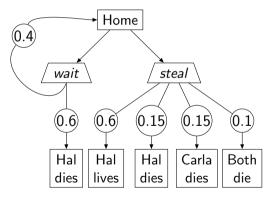


Adapted from [Coleman, 1992] and [Atkinson and Bench-Capon, 2008].

Lost Insulin Hypothetical Retrospections

The interests of Hal and Carla are in conflict. There may be *negative retrospection* (like regret) after a choice.

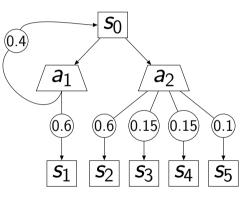
- If Hal waits at home and lives, he will be relieved!
- If Hal waits at home and dies, he will regret not taking Carla's insulin.
- If Hal steals the insulin, it he and Carla die, he may regret because Carla didn't have to die.
- If Hal steals the insulin and it does not work, he might regret breaking the law unnecessarily.



Adaptation for Planning

Multi-Moral Markov Decision Process MMMDP: $\langle S, A, P, s_0, H, M, C, L \rangle$.

- S finite set of states.
- $A = \{wait, steal\}$ finite set of actions
- $P: S \times A \times S \rightarrow [0, 1]$ probabilistic transition function
- $s_0 \in S$ initial state.
- $H \in \mathbb{N}$ horizon.
- *M* set of moral theories.
- C set of moral considerations.
- $L: M \to \mathbb{R}$ weak lexicographic ranking.



- Each Moral Consideration is a tuple, $c = \langle \mathcal{W}, J, Q, \preceq, \approx \rangle \in C.$
 - *W* represents the space of morally relevant information, or *moral worth*.
 - $J: S \times A \times S \rightarrow W$ is a judgment function.
 - Q^W: (W × [0,1])ⁿ → W aggregates moral worth given baseline worth function W : S → W.

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$\textbf{Utilitarianism} \rightarrow \textbf{Utility}$

- $\mathcal{W} = \mathbb{R}$
- J(s, a, s'): Hal or Carla's *pleasure verses* pain from a state transition.

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$$Q^{W}(W', P) = \sum_{i \in 1...|W|} P_i \cdot (w'_i + w_i)$$

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Absolute Deontology \rightarrow Duty Violation

- $\mathcal{W} = \{\top, \bot\}$
- $J(s, a, s') = \top$ if law violation, \perp otherwise.
- $Q^W(W', P) = \bigvee_{i \in |W'|} (P_i > 0 \land w'_i) \lor w_i$

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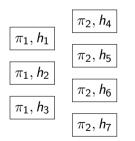
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We generally shorten the aggregation function to $\mathcal{Q}^{W}(s, a)$.

MEHR over policies

- Use Multi-Objective Heuristic Dynamic Programming [Chen et al., 2023] to find all Pareto undominated policies.
- Extract histories/trajectories from each policy, then feed into MEHR.



- Generate an argument in support of each policy from the perspective of each history Arg(π, h):
 From the initial state s₀, it was acceptable to perform policy π, resulting in consequences h with probability P(h).
- Attacks generated from two critical questions:
 CQ1: Did h' violate the moral theory and h did not?
 CQ2: Was there greater expectation that π' would violate the moral theory than π?

 $\mathsf{MMMDP} = \langle S, A, P, s_0, H, M, C, L \rangle$

Each moral theory $m = \langle C^m, \psi \rangle \in M$ defines critical questions from (a number of) moral considerations.

- $C^m \subseteq M$ are considerations relevant to the theory
- $\psi: 2^{Arg} \rightarrow \{ \leftarrow, \circ \}$ defines attacks in MEHR.

$$CQ1 = W^{h}[0](s_{0}) \succ W^{h'}[0](s_{0})$$

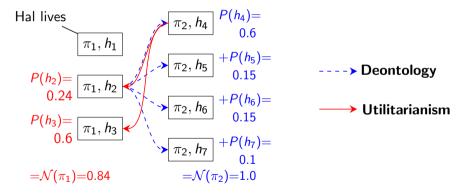
 $CQ2 = Q^{\pi}(s_{0}, \pi(s_{0}, 0)) \succ Q^{\pi'}(s_{0}, \pi'(s_{0}, 0))$

 $\Psi(\textit{Arg}(\pi,h),\textit{Arg}(\pi',h')) = \leftarrow$ if $\textit{CQ1} \land \textit{CQ2}$ otherwise \circ

Lost Insulin MEHR Graph

Selected policy has minimal *negative retrospection* over supporting arguments and moral theories.

Waiting Policy π_1 vs. Stealing Policy π_2



In this case, π_1 is preferred.

Conclusion

There's more!

- Weak Lexicographic Ordering $L: M \to \mathbb{R}$ establishes preference between theories in MEHR.
- We also have a Multi Moral Stochastic Shortest Path with non-moral cost consideration R ∈ C, budget b ∈ ℝ⁺, goal states G ⊆ S.
- Exponential # of histories in time for each policy \rightarrow exponential time/space complexity.
- Results from expanded Lost Insulin example.

Future Work:

- Performance improvements; approximation methods.
- Counterfactual explanations: 'what if a₁ on s₃?'
- More moral theories! More case studies!

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