

Legibility and Predictability of Robot Motion

Central Insight:

Predictability and legibility are fundamentally different and often contradictory properties of motion

I. Formalizing legibility and predictability

Definition 1: Legible motion is motion that enables an observer to quickly and confidently infer the correct goal G .

Definition 2: Predictable motion is motion that matches what an observer would expect, given goal G .

Legibility: Observer makes inference $I_L: \Xi \rightarrow G$

→ legible motion: confidently infer the correct goal G given a snippet of trajectory, $\xi_{s \rightarrow a}$, $a = \xi(t)$

$$I_L(\xi_{s \rightarrow a}) = G$$

Predictability: Observer knows the goal, expect certain action from the robot, $I_P: G \rightarrow \Xi$

→ predictable motion: $I_P(G) = \xi_{s \rightarrow a}$

II. Modeling Predictable Motion

• I_P : robot acts "rationally", "efficiently" & cost function.

$$C: \Xi \rightarrow \mathbb{R}^+$$

lower costs → max "efficient" trajectories: $I_P(G) = \arg \min_{\xi \in \Xi_{s \rightarrow a}} C(\xi)$

• Predictability score $\in [0, 1]$

predictability(ξ) = $\exp(-C(\xi))$ (goal: maximize this score)

III Modeling Legible Motion.

• I_L : which end state will be most "efficiently" achieved by $\xi_{s \rightarrow a}$

$$\rightarrow I_L(\xi_{s \rightarrow a}) = \arg \max_{G \in \mathcal{G}} P(G | \xi_{s \rightarrow a})$$

Using Baye's Rule: $P(G | \xi_{s \rightarrow a}) \propto P(\xi_{s \rightarrow a} | G) P(G)$

$$P(\xi_{s \rightarrow a} | G) = \frac{\int_{\xi_{a \rightarrow a}} P(\xi_{s \rightarrow a \rightarrow a})}{\int_{\xi_{a \rightarrow a}} P(\xi_{s \rightarrow a})} \rightarrow \begin{array}{l} \text{all trajectories from } s \rightarrow a \rightarrow G \\ \text{all trajectories from } s \rightarrow G \end{array}$$

$$P(\xi_{s \rightarrow a} | G) = \frac{P(\xi_{s \rightarrow a}) \int_{\xi_{a \rightarrow a}} P(\xi_{a \rightarrow a})}{\int_{\xi_{a \rightarrow a}} P(\xi_{s \rightarrow a})} \quad \text{assuming } P(\xi_{x \rightarrow y \rightarrow z}) = P(\xi_{x \rightarrow y}) P(\xi_{y \rightarrow z}) \quad \text{separable}$$

Principle of maximum entropy:

$$P(\xi_{s \rightarrow a} | G) \propto \frac{\exp(-C(\xi_{s \rightarrow a})) \int_{\xi_{s \rightarrow b}} \exp(-C(\xi_{s \rightarrow b}))}{\int_{\xi_{s \rightarrow b}} \exp(-C(\xi_{s \rightarrow b}))}$$

Using Laplace's method to approximate probabilities. If C is quadratic, its Hessian is constant

$$\rightarrow \int_{\xi_{x \rightarrow y}} \exp(-C(\xi_{x \rightarrow y})) \approx k \exp(-C(\xi_{x \rightarrow y}^*)) \quad \text{optimal trajectory}$$

$$\rightarrow P(G | \xi_{s \rightarrow a}) \propto \frac{\exp(-C(\xi_{s \rightarrow a}) - C(\xi_{s \rightarrow a}^*))}{\exp(-C(\xi_{s \rightarrow a}^*))} P(G)$$

- $\text{legibility}(\xi) = \frac{\int P(G^* | \xi_{s \rightarrow a}(t)) f(t) dt}{\int f(t) dt}$ $f(t) = \text{trajectory weight} = T - t$ ($T = \text{duration of the trajectory}$)
- To prevent the robot from going too far, add a regularizer.

$$L(\xi) = \text{legibility}(\xi) - \lambda C(\xi)$$

Implication:

Collaborative robots must be legible in all collaboration paradigms