Motion Planning and Continuous Control in a Unified Cognitive Architecture

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Cognitive Architectures

A *cognitive architecture* is a unified theory of the mind that:

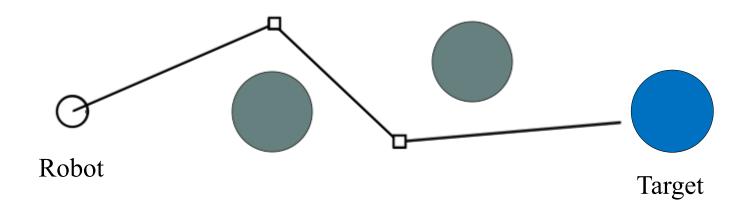
- Specifies what remains *constant* across different domains
- Incorporates many assumptions from *cognitive psychology*
- Offers a *programming language* for building intelligent systems

The PUG architecture (Langley et al., 2016) assumes that:

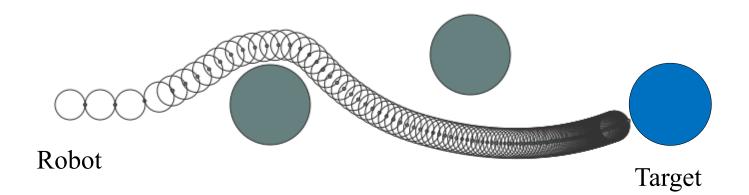
- Symbolic relations are grounded in quantitative descriptions
- Relations have associated utilities that reflect tradeoffs
- Discrete skills have associated control equations
- Mental simulation creates trajectories to guide planning

This talk reports a recent extension – PUG/C – that unifies symbolic and numeric processing more deeply.

PUG's Navigation Behavior



Desired Navigation Behavior



PUG's Knowledge Structures

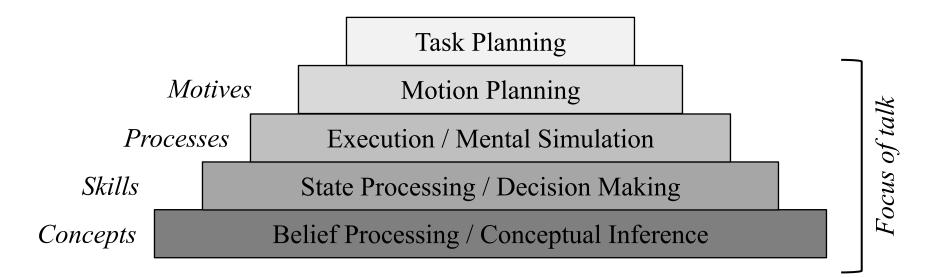
PUG/C incorporates four distinct types of generic, long-term knowledge structures:

- *Concepts* Define relational categories, attributes, and *veracity*
- *Skills* Specify *control values* based on match to *target concepts*
- **Processes** Predict changes in attributes given current values
- *Motives* Indicate *utility* of relations conditioned on situation

The architecture uses these elements for conceptual inference, reactive control, heuristic evaluation, and plan generation.

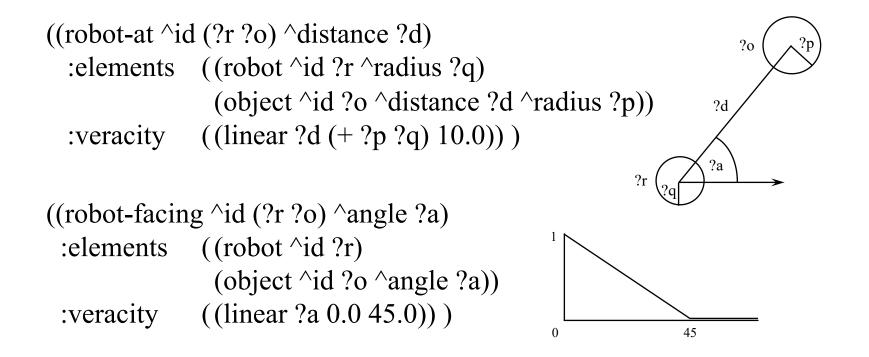
PUG's Layered Processes

Like other cognitive architectures, PUG/C operates in *cycles* that use knowledge to produce new short-term structures.



These levels are organized in a *cascaded* manner, with each one using results produced by those below it.

PUG Concepts



Here the function (*linear obs max min*) returns 1 if the observed value $obs \le max$, 0 if $obs \ge min$, and obs/(max - min) when max < obs < min.

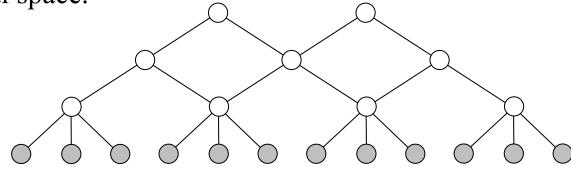
Beliefs are ground instances of concepts that relate specific entities.

Conceptual Inference in PUG

At the lowest level, *conceptual inference* derives beliefs that are consistent with perceptions / predictions:

- Matches conceptual rules to infer beliefs like (*robot-at R1 O1*)
 - Computes *values* of numeric attributes associated with beliefs
 - Calculates *veracity* (degree of match) for each inferred belief
- Applies this recursively to generate the full deductive closure

Together, the resulting beliefs describe the current *state* as a point in N-dimensional space.



PUG Beliefs

Perceptions:

Veracity Utility

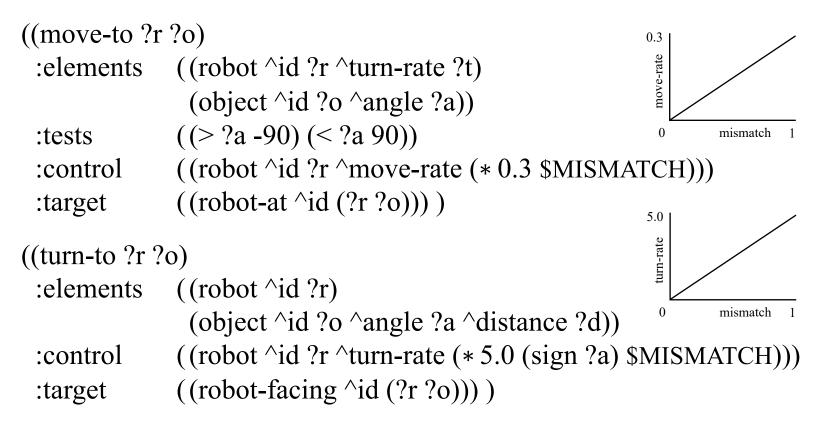
(robot ^id R1 ^radius 0.15 ^move-rate 0.0 ^turn-rate 0.0)1.000.0(object ^id O1 ^distance 2.0 ^angle 0.0 ^radius 0.4)1.000.0(object ^id O2 ^distance 4.123 ^angle 14.03 ^radius 0.4)1.000.0(object ^id O3 ^distance 6.0 ^angle 0.0 ^radius 0.4)1.000.0

Inferred Beliefs:

(robot-at ^id (R1 O1) ^distance 2.0) 0.85 0.0 (robot-at ^id (R1 O2) ^distance 4.12) 0.62 0.0 (robot-facing ^id (R1 O1) ^angle 0.0) 1.00 0.0 (robot-facing ^id (R1 O2) ^angle 14.03) 0.69 0.0 (robot-facing ^id (R1 O3) ^angle 0.0) 1.00 0.0 O2 (approaching ^id (R1 O1)) 0.91 -20.04.12 14.0

R1

PUG Skills



Here the symbol *\$MISMATCH* stands for one minus the *veracity* of the matched target concept.

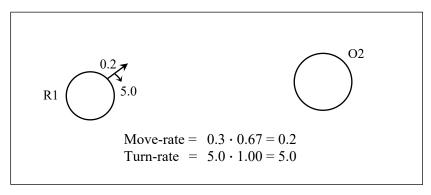
Intentions are ground instances of skills that involve specific entities.

State Processing / Decision Making

When PUG carries an out an active intention associated with skill S, whether mentally or externally, it:

- Checks that S's conditions match the current beliefs
- Finds degree of mismatch M to S's target belief
- Ensures the mismatch does not fall below threshold
- Else inserts M into S's equations to find control values

If *multiple* intentions apply, then PUG takes the *vector sum* of control values (as with potential fields).



PUG Processes

Here the functions **dd* and **da* calculate change in the robot's *distance* and *angle* relative to an object as it moves forward.

Reactive Execution / Mental Simulation

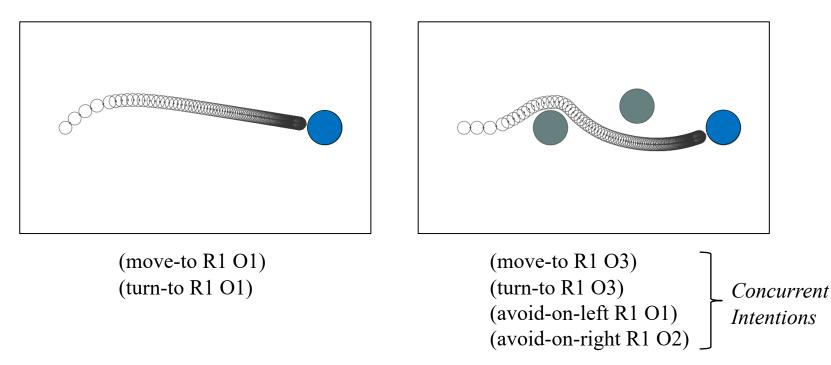
PUG/C applies skills and invokes state processing repeatedly to generate *motion trajectories*.

- When the architecture carries out actions in the environment, this corresponds to *reactive control*.
 - This operation does *not* require processes, as it has no need to predict future states.
- When the system imagines carrying out actions in its mind, it corresponds to *mental simulation*.
 - This operation *does* rely on processes (applied in parallel) to predict the succeeding state.

Each trajectory follows deterministically from an intention set (i.e., skill instances) that constitute a *motion plan*.

Reactive Execution / Mental Simulation

PUG/C applies skills and invokes state processing repeatedly to generate *motion trajectories*.



Each trajectory follows deterministically from an intention set (i.e., skill instances) that constitute a *motion plan*.

PUG Motives

```
((robot-at ^id (?r ?o))
 :conditions ((robot ^id ?r ^radius ?rr)
               (object ^id ?o ^type target ^distance ?d ^radius ?or))
 :utility (cond ((<?d (+?rr ?or 0.25)) 10.0))
                    (t \ 0.0))
             achievement)
 :type
((approaching ^id (?r ?o))
 :conditions ((robot ^id ?r ^radius ?rr)
               (object ^id ?o ^type obstacle ^distance ?d ^radius ?or))
             (cond ((<?d (+?rr ?or)) -20.0))
 :utility
                    (t \ 0.0))
             maintenance)
 :type
```

An *achievement* motive assigns utility to a belief only on its first match, where a *maintenance* motive does so repeatedly.

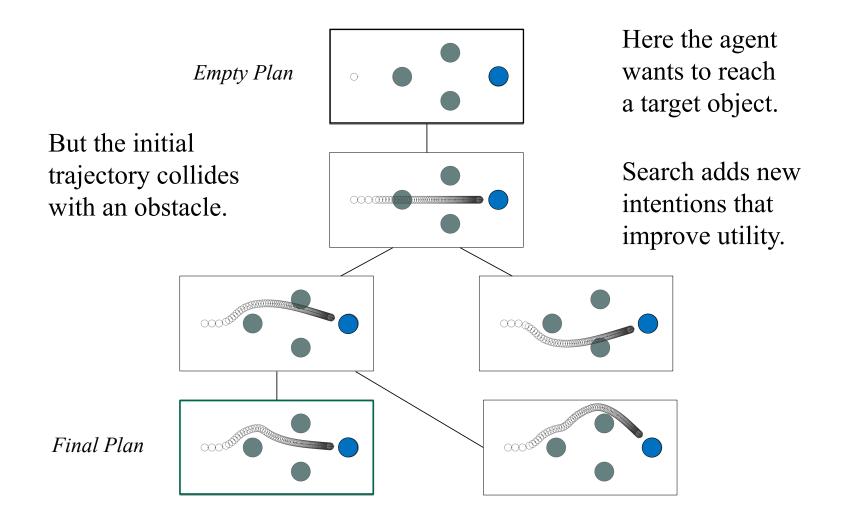
Heuristic Search for Motion Plans

PUG/C combines motives with mental simulation to carry out greedy search through a space of motion plans.

- Starting from an empty plan, it retrieves intentions that would *achieve its target beliefs*.
- Mental simulation produces a trajectory and uses motives to assign *utilities* to each belief and state.
- If some of these beliefs have *negative utility*, PUG retrieves intentions that would help *avoid* them.
- When multiple repairs are possible, it selects the candidate plan / trajectory with the *highest utility*.

This search continues until it finds an acceptable motion plan.

Heuristic Search for Motion Plans



Related and Future Research

PUG/C incorporates insights from a number of paradigms:

- Cognitive architectures (Soar, ICARUS, teleoreactive systems)
- Error-driven feedback control and potential fields
- Qualitative reasoning and quantitative simulation

But it combines ideas from these efforts in a unified framework.

Future research will add support for *places* (virtual objects) and large-scale *maps* (topological networks).

We will also integrate PUG/C with PUG/X, an earlier extension that combined *task planning*, *execution*, and *monitoring*.

Summary Remarks

PUG/C is a cognitive architecture for embodied, human-like agents that incorporates:

- Concepts, motives, skills, and processes that have both *symbolic* and *numeric* elements
- *Cascaded processing* with layers for inference, state processing, mental simulation, motion planning, and task planning
- Skills' target concepts match to different degrees, which serve as error signals that enable continuous control
- Mental simulation and motivational processing support *greedy search* through a space of continuous motion plans

We demonstrated PUG/C's behavior on scenarios that parallel skill application, obstacle avoidance, and motion planning.

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PUG's Layered Processes

Like other cognitive architectures, PUG/C operates in *cycles* that use knowledge to produce new short-term structures.

The framework differs in that it relies on five processing levels:

- *Belief processing* Inference from perceptions / predictions
- *State processing* Applies skills, processes, motives
- *Execution / Mental simulation* Generates trajectories
- *Motion planning* Heuristic search for an intention set
- *Task planning* Search for a sequence of motion plans

These levels are organized in a *cascaded* manner, with each one using results produced by those below it.