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Omniscient Debugging for GOAL Agents in Eclipse
(Demonstration)

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Abstract

The main goal of our demonstration is to show how omniscient debugging can be applied in practice to cognitive agents. A concrete implementation of the mechanisms proposed in [Koeman et al., 2017] has been created for the GOAL agent programming language in the Eclipse environment, integrated with the source-level debugger of [Koeman et al., 2016], thus fully implementing the proposal within a state-of-the-art setting. The implementation will be used together with typical agent programs to demonstrate its practical use.

1 Introduction

Running the same agent system again more often than not results in a different program run or trace, which complicates the iterative process of debugging [Bracha, 2012]. Therefore, in [Koeman et al., 2017], we propose a tracing mechanism that supports omniscient debugging for cognitive agents. This mechanism facilitates debugging by allowing a developer to move backwards in time through a program’s execution. Such a ‘time travelling debugger’ is regarded as one of the most powerful debugging tools [Zeller, 2009; Bracha, 2012]. We show in [Koeman et al., 2017] that the proposed tracing mechanism for cognitive agent programs is efficient and does not affect the runs of agent programs in the sense that the same failures can be reproduced when the mechanism is turned on and off. This means that the mechanism is fast enough to be used in practice for debugging failures in cognitive agents, opposed to existing implementations (e.g., for Java) that have a significant performance impact.

In [Koeman et al., 2017] we also introduce a trace visualization method tailored to cognitive agents based on a space-time view of the execution history [Azadmanesh and Hauswirth, 2015]. A developer can navigate this view, evaluate queries on a trace, and apply filters to it to obtain views of only the relevant parts of a trace. The approach is integrated with a source-level debugger and traces source code locations, which enables a developer to single-step through a program’s execution history, facilitating fault localization.

Prototypes of the proposed mechanisms have been implemented for the GOAL agent programming language [Hindriks, 2009], embedded in the GOAL plug-in for Eclipse [Koeman and Hindriks, 2015].

In this demonstration proposal, we will describe the omniscient debugging mechanisms that have been implemented for the GOAL agent programming plug-in in Eclipse. A typical example will be used in the demonstration to provide an insight into the practical use of these features, integrated with the source-level debugger [Koeman et al., 2016].

2 Demonstration

In [Koeman et al., 2017], we discuss how to efficiently facilitate reverting an agent to any previous state by recording its execution (i.e., tracing). This recording can be toggled on or off in the development environment. For efficient fault localization, however, it also needs to be easy for a developer to identify states in a program’s execution that are related to the failure under investigation. Moreover, a developer should not get lost in navigating between these states, but always have a sense what point in the execution s/he is evaluating and how the current state affected the execution.

Therefore, in [Koeman et al., 2017], we adapt the concept of a space-time view first developed in [Azadmanesh and Hauswirth, 2015] in the context of Java programming to cognitive agent programming. A space-time view is a table that is structured along space and time dimensions, where the rows in the table correspond to the space dimension, which is composed of the different elements in a state that are traced. The columns entail each step in the agent’s execution history.

For cognitive agents in GOAL, the elements in a space-time view (i.e., that are traced) are the agent’s events, beliefs, goals, actions, and modules. We use the corresponding signatures as the rows in the space dimension. For example, the signature print/1 in Fig. 1 represents a print action with one parameter. Each point in a trace represents a step (column) in the time dimension. Multiple space elements (signatures) can be used in a single step, e.g., evaluating a query may require accessing several beliefs and goals. The cells in our space-time view contain information about how an element was used at a particular step, which differs per type of element (e.g., a belief can be modified or inspected, an action or plan can be called and performed, a module can be

\[A \text{ short demonstration video of the tool is available at: https://www.youtube.com/watch?v=qAh7lnGtczVM}\]
entered or exited). Empty cells indicate the element was not used. To prevent cluttering the table, events (i.e., percepts or messages) are not given a row in the table, whilst for example receiving a new message does entail a separate step and thus an empty column in the table. Moreover, a single letter is used in each cell to ensure more information fits in a limited space: E entered a module, L left a module, M modified of a belief or goal, I inspected of a belief or goal, C called an action, or A action executed. We note that calling an action is different from executing an action, as the action’s precondition might fail after calling it (after which the action is not executed). An example of such a space-time view is shown in Fig. 1.

A developer can use and manipulate the space-time view in several ways. The signatures listed in the space-time view’s rows can be ordered based on type (e.g., beliefs next to beliefs), alphabetically, or by order of occurrence (the default). A developer can also apply a filter to a trace by for example selecting the signatures of interest (i.e., hiding rows) or entering a query that should hold in each state of interest (i.e., hiding columns). By clicking on any cell in the table, the debugger will reverse the agent to the state matching the column, allowing a developer to use all debugging tools (e.g., inspecting an agent’s beliefs and goals) in that specific historic state. This feature is also highlighted in Fig. 1.

We illustrate the use of these features for analysing a failure of the following example test condition associated with an agent program for the Blocks World for Teams (BW4T) [Johnson et al., 2009], which is one of the environments used for educational purposes. \( \text{goal(holding(B))}, \text{bel(atBlock(B))} \leadsto \text{done(pickUp(B))} \). This condition expresses that if the agent has the goal to hold block B, and believes it is at the block, that it should (eventually) pick up B. A failure to do so will lead to failure of the test condition (i.e., when the agent is terminated). Without an omniscient debugger, a developer would need to restart the agent, navigate to a point where the goal-belief query holds, and continue by manually stepping to try to understand why the action is not performed. With an omniscient debugger, we do not need to restart the agent, and can use the clues provided by the test condition itself to navigate to the last time that holding/1 and atBlock/1 were modified in the space-time view. We can do so either by double-clicking the corresponding cell, or, even faster, by using the filter query \( \text{goal(holding(B))}, \text{bel(atBlock(B))} \).

The goal is to facilitate locating bugs in an efficient manner, especially when dealing with very dynamic environments. An example of such an environment that we will also use in our demonstration is StarCraft [Griffioen and Plenge, 2016], i.e., we will use GOAL agent programs for StarCraft to illustrate how history-based source-level debugging facilitates fault localization for highly dynamic agent programs.

3 Conclusion

A concrete implementation of the mechanisms proposed in [Koeman et al., 2017] has been created for the GOAL agent programming language in the Eclipse environment, integrated with the source-level debugger of [Koeman et al., 2016], thus fully implementing the proposal within a state-of-the-art setting. The implementation will be used together with typical agent programs to demonstrate its practical use. This implementation and its source are publicly available[3] thus together with this demonstration providing a valuable example for the adaptation of omniscient debugging into other agent programming languages.

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[2] All (educational) agent environments are freely available at [https://github.com/eishub](https://github.com/eishub) Most of these projects include an assignment for (novice) agent programmers.

[3] See [https://goalapl.atlassian.net/wiki](https://goalapl.atlassian.net/wiki) for more information. Note, however, that the omniscient debugging features are still under development, and thus not released in the latest stable version(s) of the GOAL plug-in at the time of writing.
References


