

Towards Guided Omniscient Debugging in Education Using Pedagogical Execution Traces (PETs)

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Abstract

Educators frequently use trace-based debuggers for live classroom demonstrations. Yet, if a student’s attention drops during class, they have to fall back to watching recordings (providing a passive, non-interactive experience) or replaying the debugging session at home (lacking the instructor’s pedagogical context and verbal explanations). We introduce *Pedagogical Execution Traces (PETs)*, a concept that enriches execution traces with explanations, highlights and interactive questions. In this work-in-progress idea paper, we present the conceptual foundation of PETs as interactive learning artifacts, showing their applicability within JavaWiz, an educational trace-based graphical debugger. We explore PET authoring design goals and outline ongoing work regarding collaborative debugging scenarios and leveraging Large Language Models (LLMs) for trace annotation.

CCS Concepts: • **Software and its engineering** → **Software testing and debugging**; Dynamic analysis; • **Applied computing** → **Interactive learning environments**; • **Social and professional topics** → *CS1*; • **Human-centered computing** → *Human computer interaction (HCI)*.

Keywords: Omniscient Debugging, Execution Traces, Software Engineering Education, Program Comprehension, Interactive Learning Artifacts, Educational Visualization

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1 Introduction

Novices often struggle to form accurate mental models of program behavior and OOP concepts [8, 10, 24], a prerequisite for effective debugging. Live demonstrations [21], video recordings [14, 16], and time-traveling debuggers [28] each offer advantages on their own, yet they force educators to compromise between interactivity and pedagogical context. We address this gap with *Pedagogical Execution Traces (PETs)*.

We ground our work in the proven methodology of trace-based teaching [8] and implement our concepts within JavaWiz [28]. JavaWiz is an educational graphical time-travel debugger which is based on execution traces. An execution trace is a sequence of recorded events or program states that enable us to reconstruct the stack, heap, and statics fields (besides other trace-specific properties) at arbitrary points in time. In JavaWiz, these traces power various visualizations and enable users to step backward and forward through execution history. Currently, educators can save raw traces to disk and distribute them for students to replay at home.

PETs transform raw execution traces into guided learning artifacts by adding lightweight annotations, such as explanations, visual highlights, and interactive questions. Because these annotations are anchored to specific states and entities (e.g., a heap object or variable), they effectively clarify programming concepts. As a result, students can benefit from an instructor’s insights without requiring live guidance.

In this work-in-progress idea paper, we outline the core design goals driving our development of PETs: student interaction possibilities, low-friction authoring, and a robust, view-agnostic data model, as well as future work.

2 Motivation and Design Goals

Based on experience with JavaWiz, to effectively preserve pedagogical intent within a trace, we identified three main design goals (DGs) for developing the PET concept:

DG1: Annotations Must Support Novice Comprehension. To mirror live in-class guidance, PETs must support:

- contextual explanations where misconceptions typically arise (e.g., object aliasing or off-by-one errors).
- non-textual highlights to direct attention
- formative assessments (e.g., quizzes) that test learning outcomes and optionally gate the next execution step.

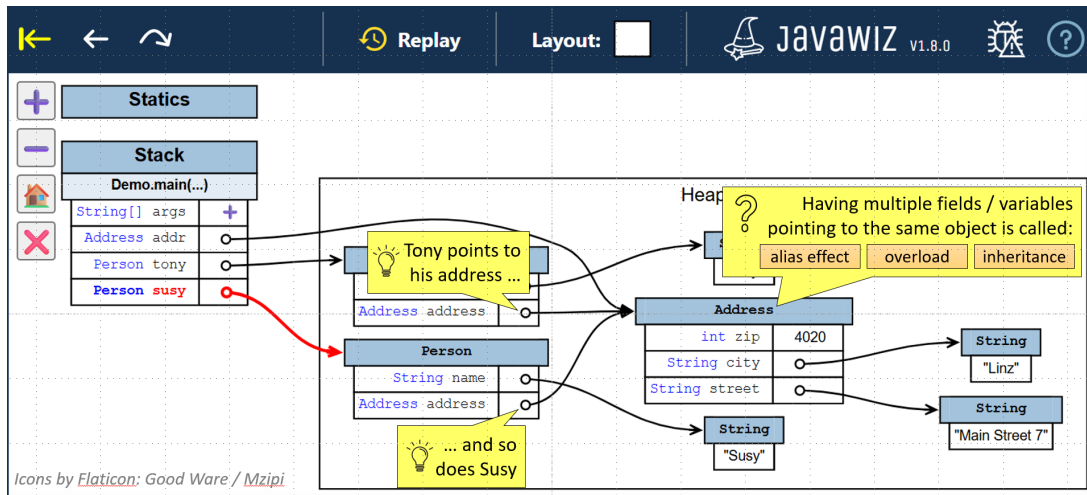


Figure 1. Mockup of JavaWiz’s annotated *Memory View*. The PET comments in line 6-13 of [Listing 1](#) translate to speech bubbles attached to field `tony.address` and field `susy.address` as well as a question linked to the heap object `tony.address`.

Listing 1. PETs are designed via “pedagogy-as-code”, i.e., via comments that reflect the instructor’s educational intent.

```

1 /* @PET Say in Controls at button StepInto: You can use
2    Step Into to inspect the constructor */
3 Address a = new Address(4020, "Linz", "Main Street 7");
4 Person tony = new Person("Tony", a);
5 Person susy = new Person("Susy", a);
6 /* @PET Say in MemoryView at field tony.address:
7    Tony points to his address... */
8 /* @PET Say in MemoryView at field susy.address:
9    ... and so does Susy */
10 /* @PET AskSingle in MemoryView at object tony.address:
11    Having multiple fields/variables pointing to
12    the same object is called
13    (*alias effect|overload|inheritance) */

```

DG2: Minimize Instructor Effort for Producing Maintainable PETs. Manual UI-based annotation is tedious and brittle. We strive for a low-friction approach by embedding the instructor’s educational intent as comments *within* the source code – we call this “*pedagogy-as-code*”. These comments compile into annotations and, crucially, survive minor code edits without requiring instructors to re-annotate.

DG3: Binding Annotations to Program Entities. Annotations must be tied to semantic program entities (e.g., objects, fields, or variables) rather than UI coordinates. Annotations must be view-agnostic (rendering consistently across different visualizations) and packaged within the trace.

3 Idea: Pedagogical Execution Traces (PETs)

We propose *Pedagogical Execution Traces (PETs)*, traces enriched with interactive *Pedagogical Annotations (PAs)* attached to individual states (or, in future work, multiple states or state transitions). We outline the conceptual foundation of PETs, structured along our three core design goals.

3.1 Interaction Possibilities (DG1)

To effectively mirror the guidance an instructor provides during a live debugging session, we distilled the required interactions into three fundamental annotation types:

- **Say (Explanations):** Contextual “speech bubbles” embedded directly within the visualizations. These bubbles are attached to specific entities (e.g., a newly allocated object in the heap or a specific method frame on the stack) at the exact moment they become relevant.
- **Highlights (Focus Points):** Visual highlights (e.g., colored borders or dimming the surrounding view) to direct the learner’s attention to a specific graphical element. This is crucial for guiding the eye during complex state changes (e.g., linked list operations)
- **Ask (Questions):** Assessments embedded within the trace. These can take the form of single-choice, multiple-choice, or short free-text questions. Crucially, “Ask” annotations can act as *execution gates*, requiring the student to predict the next state (e.g., “Which variable will change in the next step?”) before allowing them to continue stepping through the trace.

3.2 Authoring Tools (DG2)

A major adoption barrier for educational tools is authoring friction. Initially, we considered a graphical “authoring mode” within JavaWiz, where instructors could manually place annotations on the screen after recording a trace. However, coordinate-based placements break across different screen sizes, and even minor changes to the source code would require the instructor to re-record and re-annotate the entire session. Instead, we propose a “pedagogy-as-code” approach using structured source code comments that are automatically compiled into the execution trace. Comments also allow the pedagogy to be version-controlled alongside the code.

The example in Listing 1 could be visualized as Figure 1 and demonstrates how an educator can annotate code to demonstrate the concept of object aliasing. While the exact syntax is still work-in-progress, we target a declarative *What-Where-Target-When* grammar with optional parts as follows:

```
/* @PET <Action> [in <View>] [at <Target>]
   [when <Condition>]: <Payload> */
```

Action is either “Say”, “Highlight”, or “Ask”. *View* restricts the annotation to a specific visualization (e.g. the Memory View). *Target* defines at which entity the annotation should be anchored (see next section), and *Condition* (e.g., when `i == 0`) will allow instructors to annotate loops and recursive calls in more detail. This structure is designed to be highly readable for humans while keeping distraction low.

Regarding workflow, the annotated source code is an instructor authoring tool. When exported for students, @PET comments are stripped from the embedded code and stored purely as structural metadata within the trace states.

3.3 Identifying Program Entities (DG3)

PAs are anchored to semantic targets to remain stable even if the instructor adds or removes lines of code. This system also ensures that annotations are view-agnostic and layout-independent. The frontend simply queries the graphical node associated with the given target and renders the annotation accordingly. Crucially, as presented in Figure 1, we distinguish between variables (e.g., field `tony.address`) and objects (e.g., object `tony.address`). Table 1 lists entities that our initial version of PETs for JavaWiz will support.

Table 1. The different targets for Pedagogical Annotations.

<i>Group</i>	<i>Kind</i>	<i>Example</i>
Variables	Local	<i>at local x</i>
	Static	<i>at staticClazz.MAX</i>
	Array Element	<i>at element arr[4]</i>
	Fields	<i>at field tony.address</i>
Objects	Object	<i>at object tony.address</i>
Control Flow	Class	<i>at classClazz</i>
	Method	<i>at methodClazz.foo</i>
	Line	<i>at lineClazz:8</i>
JavaWiz-Specific	Buttons	<i>at buttonStepInto</i>

4 State-of-the-art and Related Work

Omniscient debuggers / *Capture-and-replay debuggers* capture full execution histories, while *time-traveling debuggers* provide retrospective backward/forward navigation [12, 17–20, 23, 26, 27]. These approaches rely on dynamic analysis over execution traces [4, 7, 13, 22] (see Cornelissen et al. [5] for a comprehensive survey). PETs leverage the same foundations but introduce a layer that encodes pedagogical intent directly into the trace.

Educational visualizers such as Jeliot [1–3, 15], jGRASP [6], Python/Java Tutor [9], Localizer [11], and various heap visualization tools [29, 30] help novices build mental models of runtime concepts (see Sorva et al. [25] for a comprehensive survey). However, while these systems visualize executions, they do not package embedded pedagogy. Building on JavaWiz [28], PETs differentiate themselves by binding explanations and questions to specific semantic entities via an easy-to-maintain “pedagogy-as-code” authoring model.

5 Ongoing and Future Work

PETs open several avenues for future research, ranging from pedagogical evaluations to advanced AI integrations.

5.1 Student Interaction Styles

To measure the pedagogical impact of PETs, we plan to conduct user studies comparing cognitive load, engagement, and retention across three scenarios: live instructor presentations, passive video recordings, and active PET stepping. We hypothesize that PETs’ self-paced nature and “Ask” execution gates will foster active learning and increase retention.

5.2 Collaborative Educational Scenarios

PETs enable “Debug With Me” sessions: students follow an instructor’s debugging session live on their own devices and place annotations to answer instructor prompts (e.g., predicting memory changes). The instructor can later review class-annotated PETs to gauge student engagement / comprehension. Further, having students author PET comments themselves as homework (or collaboratively during live coding) turns the annotation process itself into an active pedagogical exercise in predicting program behavior [8].

5.3 AI-Assisted and Autonomous Debugging

To further reduce authoring friction, we investigate LLMs for automated PET generation. We currently conduct a study on optimizing LLM inputs (source code vs. raw trace vs. images) to maximize explanation quality while minimizing latency. Additionally, we compare student ratings of AI-generated versus human-authored PETs. We further envision stronger LLM integration into JavaWiz via the Model Context Protocol (MCP). Rather than statically generating PETs, an LLM could then autonomously run JavaWiz, dynamically step through code, open relevant visualizations, and inject real-time annotations based on perceived student struggles, effectively acting as a personalized, interactive tutor.

6 Conclusion

Traditional debuggers are powerful diagnostic tools, but they lack the pedagogical context required for effective software engineering education. While live demonstrations provide this context, they are ephemeral; while video recordings preserve it, they sacrifice interactivity.

In this work-in-progress idea paper, we introduced the concept of *Pedagogical Execution Traces (PETs)* to bridge this gap. Extending the trace-based debugging capabilities of JavaWiz, PETs transform execution traces into reusable, self-paced learning artifacts. We outlined a “pedagogy-as-code” authoring approach that allows instructors to express explanations, highlights, and questions via source code comments.

Looking forward, the integration of LLMs for automated annotation and the development of collaborative, in-class debugging scenarios represent exciting frontiers. Ultimately, we envision that concepts such as our PETs can help debuggers to evolve beyond mere diagnostic utilities into active, intelligent partners in the learning process.

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