

Code Shaping: Iterative Code Editing with Free-form Sketching

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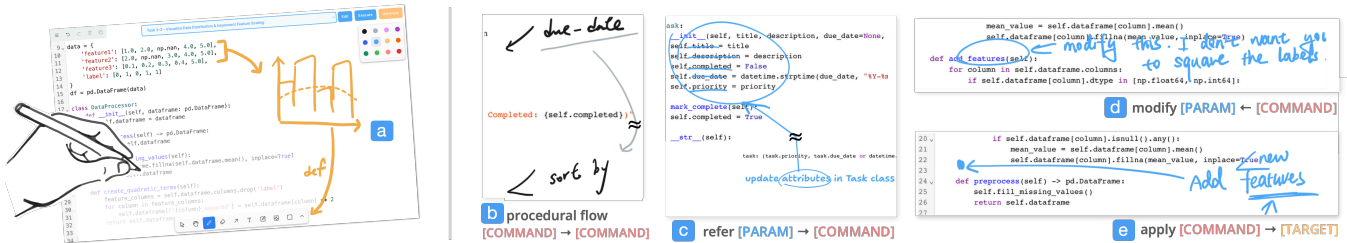


Figure 1: (a) A programmer sketches an arrow pointing from data attributes to a drawn bar chart and annotates the code with *def* to generate edited code. Right: Data collected from the user study showing how users employ arrows (\rightarrow) for different purposes, including command (the intended action of operation), parameter (supplementing the command), and target (the area where the edit should occur); (b) indicating procedural flow between commands; (c) referring to data attributes; (d) modifying a function, with the function as the parameter to supplement the command; (e) applying changes to a target area.

ABSTRACT

We present an initial step towards building a system for programmers to edit code using free-form sketch annotations drawn directly onto editor and output windows. Using a working prototype system as a technical probe, an exploratory study ($N = 6$) examines how programmers sketch to annotate Python code to communicate edits for an AI model to perform. The results reveal personalized workflow strategies and how similar annotations vary in abstractness and intention across different scenarios and users.

CCS CONCEPTS

• User interface programming; • Interaction techniques;

KEYWORDS

ink-based sketching, programming interface

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1 INTRODUCTION

Many programmers use free-form sketching to externalize ideas to plan high-level structure, workout algorithms, and annotate code. Code annotations in particular serve various purposes, such as

enhancing code comprehension [2, 12], communicating with collaborators [6], and planning future edits [10, 11]. However, past practice and previous research mostly treat sketched annotations on code as static externalizations of a programmer’s thoughts [8], not as actionable commands to interactively edit code. We propose a sketch-based editing approach where a programmer iteratively draws free-form annotations on and around a code editor to iteratively modify structure, flow, and syntax: a concept we call *code shaping*. For example, to insert a new function to draw charts, the programmer could circle lines of code about data attributes, draw an arrow to a sketch of a graph, then draw an arrow with the word “def” leading back to an insertion point in the code (Figure 1a).

We differentiate the concept of code shaping from another line of research that focuses on converting sketched drawings, such as visualizations or user interfaces on a canvas, into code [5, 13]. Those sketches directly represent the final output, without considering the syntactic structure of the code. Instead, our study explored sketches on code editors. These annotated sketches encapsulated programmers’ expectations of how the program would run and connected these sketches to the syntax code. While recent advances in multi-modal large language models have made this concept more feasible, this approach introduces challenges that need to be understood and addressed. First, similar annotations could have different meanings across various scenarios and tasks (Figure 1b-e), stemming from the ambiguous nature of sketches [1, 3]. Second, the obscure nature of AI models forces users to guess the reasons for recognition failures and opportunistically change the instruction to make them work [4, 14]. These challenges highlight the need for a system where users can iteratively clarify and modify both code and annotations. We conducted an exploratory study using a prototype code-shaping system that transforms free-form sketches on a code editor into actual code edits.

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2 PROTOTYPE SYSTEM

Our system integrates a code editor within a canvas environment to enable code shaping. The interface supports various free-form sketching tools, including colour selectors, pens, erasers, and shapes. A text tool is available for conventional editing. Two-finger panning and zooming navigate the code in the editor. A pointer tool can select one or more annotations. Pressing a “Generate” button uses all current annotations, or only those selected, as parameters for generating edited code. GPT-4o is used to recognize free-form annotations and generate code edits. To do this, syntax colour highlighting is removed from the code editor, the HTML is converted to SVG, coordinates are added, and handwritten text is recognized. The system then considers the annotations alongside previous iterations of sketch editing and the relevant codebase as part of the input context for code generation. After the code is generated, a diffing algorithm is employed to visualize which sections of code changed [7]. The user can press a “Run” button to execute the code, with text or image results shown in the console panel underneath. Users also can annotate executed results as part of their sketches.

3 EXPLORATORY STUDY

We recruited six participants (1 left-handed), aged 23 to 28, with 4 identifying as women and 2 as men. Participants were recruited through convenience sampling and received \$30 for completing the study. All participants had 2-8 years of programming experience and had used ChatGPT or Copilot 3-12 times per week. We designed three coding scenarios, each with two tasks that required specific edits to reach the goal for each scenario. These scenarios covered basic Python with object-oriented programming, machine learning with functional programming, and data engineering with declarative programming. Starter code was provided for each task, requiring edits in more than two areas. For example, a task was to extend a class to handle data points with categorical features, requiring changes to current methods to encode features and modify distance calculations accordingly. All tasks were pre-tested to ensure that GPT-4o could not generate the correct code immediately. Participants were assigned 2 out of 3 scenarios that they were more familiar with based on a pre-screening questionnaire. They completed 2 SCENARIOS × 2 TASKS each in 50 minutes. Followed by a post-study 7-point Likert scale questionnaire and a semi-structured interview. System logs, screen recordings, and interview scripts were collected for analysis.

4 RESULTS

All but two participants completed the four assigned tasks, the two participants failed to complete one task within the assigned time. The clarity of the effect of their sketches on generated code ($Mdn = 3.5, SD = 1.83$) and the ease of iterating on sketches ($Mdn = 4, SD = 2.34$) was rated low. This aligns with our interview results, where participants noted an unclear mapping between sketches and edited code, especially with multiple annotations. Four participants expressed a desire to see their annotations “stick” to the corresponding changed code segments.

Personalized Workflow. We observed participants gradually develop a personalized workflow for editing code with sketches.

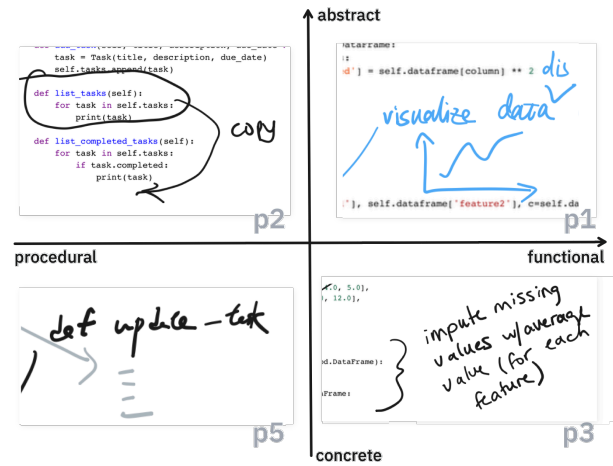


Figure 2: The classification of sketched annotations from participants situated in a quadrant with two dimensions, ABSTRACT-CONCRETE and PROCEDURAL-FUNCTIONAL.

P2 found that breaking down tasks into very low-level details was ineffective for AI interpretation, while P5 emphasized the need for smaller task pieces for better system understanding. Participants sometimes wrote higher-level instructions first when unsure about the solution, but had a rough idea of where the code edits should happen and what the “*shape of the code looks like*” [P4]. After evaluating the generated code, they then added annotations for lower-level code editing based on their approaches in mind.

Types of Sketches. Participants used similar sketches for different purposes, such as arrows pointing to context [P1] or targets of changes [P4] (Figure 1b-e). Overall, the sketches could be situated in a quadrant with two dimensions (Figure 2). ABSTRACT-CONCRETE describes whether the annotations are abstract symbols or graphs versus concrete written text. PROCEDURAL-FUNCTIONAL classifies the target of the annotations, ranging from procedural steps describing how the program should be structured and run to functional descriptions specifying how the program should work. Participants often combined these aspects, drawing graphs and adding arrows to refer to certain data attributes, specifying both functional and procedural terms.

Sketch as a Tool. All participants considered sketches more than static digital ink drawings, also treating them as functional “tools” that could be reused [9]. They expressed how they could use different sketches to achieve the same effect, choosing which sketch to use based on the environment, such as available white spaces. They also reused their sketches to convey the same effect; for instance, an arrow used to add a function to a target code section was reused by P3 to add another function.

5 FUTURE WORK.

Currently, the user experience is hampered by both model-related interpretation errors and interaction techniques, and we plan to iterate on the design to understand the idea of code shaping further.

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