Programming as collaborative reference (full presentation)

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We argue that programming-language theory should face the pragmatic fact that humans develop software by interacting with computers in real time. This interaction not only relies on but also bears on the core design and tools of programming languages, so it should not be left to Eclipse plug-ins and StackOverflow. We further illustrate how this interaction could be improved by drawing from existing research on natural-language dialogue, specifically on collaborative reference.

The users and implementations of modern programming languages are stymied by a communication bottleneck: Programs are so long that the need to express them in full detail imposes a heavy cognitive and physical burden even if one already understands the exact algorithm intended. Whenever you’re programming and struggle to enter a well-formed programs in today’s programming languages want takes expression even though it would be obvious to another programmer looking over your shoulder what you mean, that’s an instance of this bottleneck in the human-to-computer direction.

It would be excusable if the reason the program we want takes \( n \) bits to express is that there are \( 2^n - 1 \) other programs that we might want. But the vast majority of well-formed programs in today’s programming languages are undesirable, and many desirable programs are equivalent. If only we can just input a desired program by entering the number of lexicographically preceding desired programs! But that’s intractable for humans and computers alike. Abstractions and types may reduce this communication burden somewhat, but the boilerplate necessary to access highly parametrized abstractions and the annotations necessary to help type inference along contribute to the problem themselves.

We face a similar problem when talking to fellow human beings. Most of the meanings we intend to convey using natural-language utterances would be excruciating to spell out fully in any language. For example, pronouns (such as “him”) and descriptions (“the president”) are on one hand much more ambiguous than proper names (“Barack Obama”) and explicit variable names (“\( x \)”), but on the other hand more concise as well as easier to process for speakers and hearers alike. To take another example, the familiar question “Can everyone hear me?” is typically spoken to ask about everyone in the room rather than everyone in the world \([4]\). It scantily helps to rephrase the question to “Can everyone in the room hear me?”—Which room? (Do you mean the city of Philadelphia or the movie Philadelphia?)

Despite such odds, humans manage to communicate all the time—whether to name a room, a proof, or a program. Key to this feat are

- our use of context, which narrows down the meanings that make sense to convey, and
- our exchange of feedback, which makes it unnecessary to get an utterance right the first time or to ever give a complete explicit description.

Programming languages today already use some context to lessen the burden of communication; examples include scope, type inference, and overloading resolution. Development environments today also provide feedback in rudimentary forms such as continuous compilation and identifier completion. Comparing these facilities against human communication shows that it is possible for us to program more concisely, especially if the language is designed with context and feedback in mind. And program more concisely we must, as we build larger systems and check stronger properties, which might be expressed today in fancy type systems that require many (often “obvious”) annotations.

For example, whenever you want to talk to me about something or somebody, maybe someone you spot across the stadium, we usually manage to agree on the same person pretty quickly, even though we haven’t agreed on a numbering of all objects ahead of time. The psychologists Clark and Wilkes-Gibbs \([2]\) point out that this task is very interactive—not at all like taking one shot to write out a clear noun phrase.

A: the guy reading with, holding his book to the left.
B: Okay, kind of standing up?
A: Yeah.
B: Okay.

Clark and Wilkes-Gibbs study this collaborative reference task by asking experiment subjects to perform a matching task in pairs. Part of their study classified the moves people tend to make and identified how they signal and anticipate these moves.

Programming is a human-computer collaboration whose goal is also reference. We are not alone to suggest easing this daunting goal by making the object of reference not just the raw code but also what it does and why it works. A flexible interaction can take advantage of how few programs make sense to relieve humans from reading and writing more boilerplate and annotations. Perhaps, what a program does would be expressed roughly by a type, and why it works would be expressed by a type derivation; like Conor McBride, we “think of types as warping our gravity, so that the direction we need to travel becomes ‘downhill’.”

What’s tricky is that the intended program is seldom the exact minimum. For example, given a signature like Haskell’s `Show` or `Ord`, sometimes the implementation you want is not the default instance that type classes give you,
but almost! So close, yet so far to have to compose all
the functor invocations by hand! So the ideal trip to the
desired program is like

- Marble Madness;
- filling in holes in Agda;
- IntelliSense, not only constrained by types, but also
  liberated by completing complex expressions;
- input method editors; and
- pair programming or code review, which goes faster
  and more smoothly as participants develop a shared
  vocabulary like between Clark and Wilkes-Gibbs’s
director and matcher.

It is crucial to allow decisions to be made (facets to be
specified) in many different ways during the collaboration.
That’s not all. We wouldn’t be here if we just wanted to
suggest that HCI researchers and IDE builders take more
context into account in the interfaces they are already
building. Neither do we want to program computers using
an existing natural language [1].

It turns out that implementing collaborative reference,
whether in natural languages or in programming lan-
guages, requires some tools familiar to you such as seman-
tics and logic:

- Participants in collaborative reference maintain
  their discourse context by updating a virtual score-
  board with constraints, or logical statements about
  the intended referent.
- The task often spawns sub/meta-tasks to clear up
collateral ambiguity, whether about a constituent
  subroutine, a constraint predicate, or a discourse
  move.
- Previously established referents are added to a vo-
cabulary that, unsurprisingly, looks a lot like a type
  environment.
- The pervasive uncertainty can be expressed using
  nondeterministic programming and managed using
  compositional heuristics for search through proba-
bility distributions with the programmer’s guidance.

DeVault’s COREF dialogue agent [3] can play both
the director role and the matcher role in a task like
Clark and Wilkes-Gibbs’s experiment task. To show
concretely that the approach is viable, we will play a
one-minute video of COREF in action (http://www.
cs.rutgers.edu/~ccshan/devault-contribution/
example-coref-matches.mp4).

An initial attempt at programming by collaborative
reference may target the problem of (overlapping) over-
loading resolution or (undecidable) type inference. The
key is to identify a space of possible discourse moves,
including the space of programs that make sense and
a variety of strategies that both the director and the
matcher can use to describe a candidate they have in
mind and distinguish it telegraphically from competitors
in play. For example, instead of insisting that all overload-
ing be unambiguous, a programming language can resolve
overloading interactively if there is enough doubt as to
which referent the programmer intends. The referent that
results from this collaboration would then become part
of the program text (in particular, part of the indexical
content [5] that any copy-and-paste operation would op-
erate on), so meta-theorems such as type safety would
still hold.

[1] Allen, James F., Nathanael Chambers, George Ferguson,
Lucian Galescu, Hyunchul Jung, Mary D. Swift, and
William Taysom. 2007. PLOW: A collaborative task learn-
ing agent. In AAAI-2007: Proceedings of the 22nd national
conference on artificial intelligence, 1514–1519. The Ameri-
can Association for Artificial Intelligence, Menlo Park, CA:
AAAI Press.
Rutgers University.
Ph.D. thesis, Department of Linguistics, University of
Massachusetts.
semantics, logic, metaphysics, and epistemology of demonstra-
tives and other indexicals. In Themes from Kaplan,
ed. Joseph Almog, John Perry, and Howard Wettstein,