Design and implementaiton of a live coding environment

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Abstract

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1 Introduction

One of the aspects that make spreadsheet tools such as Excel more accessible than programming environments is their liveness. When you change a value in a cell in Excel, the whole spreadsheet is updated instantly and you see the new results immediately.

Increasing number of programming environments aim to provide the same live experience for more standard programming languages, but doing this is not easy. Fully recomputing the whole program after every single change is inefficient and calculating how a change in source code changes the result is extremely hard when the editor allows arbitrary manipulation of program text. For example, consider the following simple program that gets the release years of 10 most expensive movies in a data set movies:

```
let top = movies

.sortBy(\lambda x \rightarrow x.getBudget()).take(10)

.map(\lambda x \rightarrow x.getReleased().format("yyyy"))
```

A live coding environment shows a preview of the list of dates top and the programmer then changes the code by making 10 a variable and changing the date format to see the full date:

```
let count = 10
let top = movies
.sortBy(\lambda x \rightarrow x.getBudget()).take(count)
.map(\lambda x \rightarrow x.getReleased().format("dd-mm-yyyy"))
```

Ideally, the live coding environment should understand the change, reuse a cached result of the first two transformations (sorting and taking 10 elements) and only evaluate the last map and format the release dates of already computed top 10 movies.

This is not difficult to achieve if we represent the program in a structured way and allow the user to edit code only via well-understood primitive operations such as "extract variable" (which has no effect on the result) or "change constant value" (which forces recomputation of subsequent transformations). However, many programmers still prefer to edit programs as free form text. In this paper, we present the design and implementation of a live coding system that is capable of reusing previously evaluated expressions as in the example above, yet, is integrated into an ordinary text editor.

The motivation for this paper is twofold. First, implementing a live programming system requires a different way of thinking about compilers and interpreters than the one presented in classic programming language literature. Second, an increasing number of programming environments incorporate aspects of live programming [X,Y], yet relatively little has been written on their architecture and implementation.

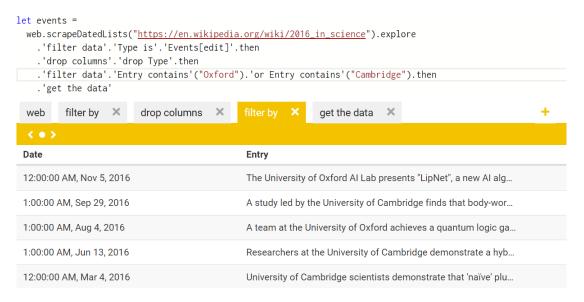
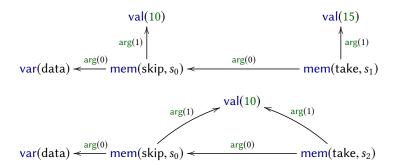


Figure 1. xxx

2 Live programming for data exploration

The Gamma introduction and what this tries to do. See pretty screenshot.

This is what we care about - it's not live programming for your typical haskell, but for R or Python.



a.) The first graph is constructed from the following initial expression:

$$let x = 15 in data.skip(10).take(x)$$

b.) The second diagram shows the updated graph after the programmer changes *x* to 10:

```
let x = 10 in data.skip(10).take(x)
```

Figure 2. Dependency graphs formed by two steps of the live programming process.

3 Formalising live coding environment

In this section, we present a formalisation of a live coding environment for a small, expression-based programming language that supports let binding, member invocations and λ abstractions. This is the necessary minimum for data exploration as described in the previous section.

It excludes constructs such as a mechanism for defining new objects as we assume that those are imported from the context through a mechanims such as type providers.

$$e = |\det x = e \text{ in } e \mid \lambda x \rightarrow e \mid e.m(e,...,e) \mid x \mid n$$

Here, m ranges over member names, x over variables and n over primitive values such as numbers. Function values can be passed as arguments to methods (provided by a type provider), but for the purpose of this paper, we do not need to be able to invoke them directly.

The problem with functions. In the context of live programming, let binding and member access are unproblematic. We can evaluate them and provide live preview for both of them, including all their sub-expressions. Function values are more problematic, because their sub-expressions cannot be evaluated. For example:

let page =
$$\lambda x \rightarrow \text{movies.skip}(x * 10).\text{take}(10)$$

We can provide live preview for the movies sub-expression, but not for movies.skip(x*10) because we cannot obtain the value of x without running the rest of the program and analysing how the function is called later.

The method described in this paper does not provide live preview for sub-expressions that contain free variables (which are not global objects provided by a type provider), but we describe possible ways of doing so in Section X and more speculative design of live coding friendly fucntions in Section Y.

3.1 Maintaining dependency graph

The key idea behind our implementation is to maintain a dependency graph with nodes representing individual operations of the computation that can be partially evaluated to obtain a preview. Each time the program text is modified, we parse it afresh (using an error-recovering parser) and bind the abstract syntax tree to the dependency graph.

We remember the previously created nodes of the graph. When binding a new expression to the graph, we reuse previously created nodes that have the same dependencies. For expressions that have a new structure, we create new nodes (using a fresh symbol to identify them).

The nodes of the graph serve as unique keys into a lookup table with previously evaluated operations of the computation. When a preview is requested, we use the node bound to the expression to find a preview, or evaluate it by first forcing the evaluation of all parents in the dependency graph.

Elements of the graph. The nodes of the graph represent individual operations to be computed. In our design, the nodes themselves are used as keys, so we attach a unique *symbol* to some of the nodes. That way, we can create two unique nodes representing, for example, access to a member named take which differ in their dependencies.

Furthermore, the graph edges are labelled with labels indicating the kind of dependency. For a method call, the labels are "first argument", "second argument" and so on. Formally:

```
s \in Symbol

i \in Integer

n \in Primitive values

x \in Variable names

m \in Member names

v \in val(n) | var(x) | mem(m, s) | fun(x, s) (Vertices)

l \in body | arg(i) (Edge labels)
```

The val node represents a primitive value and contains the value itself. Two occurences of 10 in the source code will be represented by the same node. Member access mem contains the member name, together with a unique symbol – two member access nodes with different dependencies will contain a different symbol. Dependencies of member access are labelled with arg indicating the index of the argument (the instance has index 0 and arguments start with 1).

```
\operatorname{bind}_{\Gamma,\Delta}(e_0.m(e_1,\ldots,e_n)) =
                                                                                                                    \operatorname{bind}_{\Gamma,\Lambda}(n) = \operatorname{val}(n), (\{\operatorname{val}(n)\},\emptyset)
                                                                                                                                                                                                     (4)
         v, (\{v\} \cup V_0 \cup \ldots \cup V_n, E \cup E_0 \cup \ldots \cup E_n)
                                                                                                                    \operatorname{bind}_{\Gamma,\Lambda}(x) = v, (\{v\}, \emptyset) \quad \text{when } v = \Gamma(x)
                                                                                                                                                                                                     (6)
    when v_i, (V_i, E_i) = \text{bind}_{\Gamma, \Lambda}(e_i)
    and v = \Delta(\text{mem}(m), [(v_0, \arg(0)), \dots, (v_n, \arg(n))])
                                                                                                                    \mathsf{bind}_{\Gamma,\Delta}(\lambda x \to e) = \upsilon, (\{\upsilon\} \cup V, \{e\} \cup E)
                                                                                                                                                                                                    (7)
    let E = \{(v, v_0, \arg(0)), \dots, (v, v_n, \arg(n))\}\
                                                                                                                         when \Gamma_1 = \Gamma \cup \{x, \operatorname{var}(x)\}\
\operatorname{bind}_{\Gamma,\Delta}(e_0.m(e_1,\ldots,e_n)) =
                                                                                                         (2)
                                                                                                                         and v_0, (V, E) = \text{bind}_{\Gamma_1, \Delta}(e)
         v, (\{v\} \cup V_0 \cup \ldots \cup V_n, E \cup E_0 \cup \ldots \cup E_n)
                                                                                                                         and v = \Delta(\operatorname{fun}(x), [(v_0, \operatorname{body})])
    when v_i, (V_i, E_i) = \text{bind}_{\Gamma, \Lambda}(e_i)
                                                                                                                         let e = (v, v_0, body)
    and (\text{mem}(m), [(v_0, \text{arg}(0)), \dots, (v_n, \text{arg}(n))]) \notin \text{dom}(\Delta)
                                                                                                                    \mathsf{bind}_{\Gamma,\Delta}(\lambda x \to e) = \upsilon, (\{\upsilon\} \cup V, \{e\} \cup E)
    let v = mem(m, s), s fresh
    let E = \{(v, v_0, \arg(0)), \dots, (v, v_n, \arg(n))\}
                                                                                                                         when \Gamma_1 = \Gamma \cup \{x, \operatorname{var}(x)\}\
                                                                                                                         and v_0, (V, E) = \text{bind}_{\Gamma_1, \Lambda}(e)
bind_{\Gamma,\Lambda}(let \ x = e_1 \ in \ e_2) = v, (\{v\} \cup V \cup V_1, E \cup E_1)
                                                                                                         (3)
                                                                                                                         and (\operatorname{fun}(x), [(v_0, \operatorname{body})]) \notin \operatorname{dom}(\Delta)
    let v_1, (V_1, E_1) = \operatorname{bind}_{\Gamma, \Delta}(e_1)
                                                                                                                        let v = \text{fun}(s, x), s fresh
    let \Gamma_1 = \Gamma \cup \{(x, v_1)\}
                                                                                                                         let e = (v, v_0, \text{body})
    let v, (V, E) = bind_{\Gamma_1, \Lambda}(e_2)
```

Figure 3. Rules of the binding process, which constructs a dependency graph for an expression.

Finally, nodes fun and var represent function values and variables bound by λ abstraction. For simplicity, we use variable names rather than de Bruijn indices and so renaming a bound variable foces recomputation.

Example graph. Figure 2 illustrates how we construct and update the dependency graph. Node representing take(x) depends on the argument – the number 15 – and the instance, which is a node representing skip(10). This, in turn, depends on the instance data and the number 10. Note that variables bound via let binding such as x do not appear as var nodes. The node using it depends directly on the node representing the result of the expression that is assigned to x.

After changing the value of x, we create a new graph. The dependencies of the node mem(skip, s_0) are unchanged and so the node is reused. This means that this part of the program is not recomputed. The arg(1) dependency of the take call changed and so we create a node mem(skip, s_2) with a new fresh symbol s_2 . The preview for this node is then recomputed as needed using the already known values of its dependencies.

Reusing graph nodes. The binding process takes an expression and constructs a dependency graph, reusing existing nodes when possible. For this, we keep a lookup table of member access and function value nodes. The key is formed by a node kind (for disambiguation) together with a list of dependencies. A node kind is a member access or a function:

```
k \in \operatorname{fun}(x) \mid \operatorname{mem}(m) (Node kinds)
```

Given a lookup table Δ , we write $\Delta(k, [(n_1, l_1), \ldots, (v_n, l_n)])$ to perform a lookup for a node of a kind k that has dependencies v_1, \ldots, v_n labelled with labels l_1, \ldots, l_n .

For example, when creating the graph in Figure 2 (b), we perform the following lookup for the skip member access:

```
\Delta(\text{mem}(\text{skip}), [(\text{var}(\text{data}), \text{arg}(0)), (\text{val}(10), \text{arg}(1))])
```

The lookup returns the node mem(skip, s_0) known from the previous step. We then perform the following lookup for the take member access:

```
\Delta(\text{mem}(\text{take}), \lceil (\text{mem}(\text{skip}, s_0), \text{arg}(0)), (\text{val}(10), \text{arg}(1)) \rceil)
```

In the previous graph, the argument of take was 15 rather than 10 and so this lookup fails. We then construct a new node mem(take, s_2) using a fresh symbol s_2 .

3.2 Binding an expressions to a graph

When constructing the dependency graph, our implementation annotates the nodes of the abstract syntax tree with the nodes of the dependency graph, forming a mapping $e \to v$. For this reason, we call the process *binding*.

The process of binding is defined by the rules in Figure 3. The bind function is annotated with a lookup table Δ discussed in Section 3.1 and a variable context Γ . The variable context is a map from variable names to dependency graph nodes and is used for variables bound using let binding.

When applied on an expression e, binding bind $_{\Gamma,\Delta}(e)$ returns a dependency graph (V,E) paired with a node v corresponding to the expression e. In the graph, V is a set of nodes v and E is a set of labelled edges (v_1, v_2, l) . We attach the label directly to the edge rather than keeping a separate colouring function as this makes the formalisation simpler.

Binding member access. In all rules, we recursively bind sub-expressions to get a dependency graph for each sub-expression and a graph node that represents it. The nodes representing sub-expressions are then used as dependencies for lookup into Δ , together with their labels. When binding

a member access, we reuse an existing node if it is defined by Δ (1) or we create a new node containing a fresh symbol when the domain of Δ does not contain a key describing the current member access (2).

Binding let binding. For let binding (3), we first bind the expression e_1 assigned to the variable to obtain a graph node v_1 . We then bind the body expression e_2 , but using a variable context Γ₁ that maps the value of the variable to the graph node v_1 . The variable context is used when binding a variable (6) and so all variables declared using let binding will be bound to a graph node representing the value assinged to the variable. The node bound to the overall let expression is then the graph node bound to the body expression.

Binding function values. If a function value uses its argument, we will not be able to evaluate its body. In this case, the graph node bound to a function will depend on a synthetic node var(x) that represents the variable with no value. When binding a function, we create the synthetic variable and add it to the variable context Γ₁ before binding the body. As with member access, the node representing a function may (7) or may not (8) be already present in the lookup table.

3.3 Edit and rebind loop

The binding process formalised in Section 3.2 specifies how to update the dependency graph after updated program text is parsed. During live coding, this is done repeatedly as the programmer edits code. Throughout the process, we maintain a series of lookup table states $\Delta_0, \Delta_1, \Delta_2, \ldots$ Initially, the lookup table is empty, i.e. $\Delta_0 = \emptyset$.

At a step i, we parse an expression e_i and calculate the new dependency graph and a node bound to the top-level expression using the previous Δ :

$$v, (V, E) = \mathsf{bind}_{\emptyset, \Delta_{i-1}}(e_i)$$

The new state of the node cache is then computed by adding newly created nodes from the graph (V, E) to the previous cache Δ_{i-1} . This is done for function and member nodes that contain unique symbols as defined in Figure 4. We do not need to cache nodes representing primitive values and variables as those do not contain symbols and will remain the same due to the way they are constructed.

```
\Delta_{i}(\operatorname{mem}(m), [(v_{0}, \operatorname{arg}(0)), \dots, (v_{n}, \operatorname{arg}(n))]) = v
for all \operatorname{mem}(m, s) \in V
such that (\operatorname{mem}(m, s), v_{i}, \operatorname{arg}(i)) \in E for i \in 0, \dots, n
\Delta_{i}(\operatorname{fun}(x), [(v_{1}, \operatorname{body})]) = v
for all \operatorname{fun}(x, s) \in V
such that (\operatorname{fun}(x, s), v_{1}, \operatorname{body}) \in E
\Delta_{i}(v) = \Delta_{i-1}(v) \quad (\text{otherwise})
```

Figure 4. Updating the node cache after binding a new graph

4 Evaluating previews

The mechanism for constructing dependency graphs defined in Section 3 makes it possible to provide live previews when editing code without recomputing the whole program each time the source code changes.

The nodes in the dependency graph correspond to individual operations that will be performed when running the program. When the dependencies of an operation do not change while editing code, the subsequent dependency graph will reuse a node used to represent the operation.

Our live editor keeps a map from graph nodes to live previews, so a new preview only needs to be computed when a new node appears in the dependency graph (and the user moves the cursor to a code location that corresponds to the node). This section describes how previews are evaluated.

Previews and delayed previews. As discussed in Section 3, the body of a function cannot be easily evaluated to a value if it uses the bound variable. We do not attempt to "guess" possible arguments and, instead, provide a full preview only for sub-expressions with free variables bound by a let binding. For a function body that uses the bound variable, we obtain a delayed preview, which is an expression annotated with a list of variables that need to be provided before the expression can be evaluated. We use the following notation:

```
\begin{array}{lll} p & \in & n \mid \lambda x \rightarrow e & & (Fully \ evaluated \ previews) \\ d & \in & p \mid \llbracket e \rrbracket_{\Gamma} & & (Evaluated \ and \ delayed \ previews) \end{array}
```

Here, p ranges over fully evaluated values. It can be either a primitive value (such as number, string or an object) or a function value with no free variables. A possibly delayed preview d can then be either evaluated preview p or an expression e that requires variables p. For simplicity, we use an untyped language and so p is a list of variables p and p is a list of variables p is a list of variables p in p

Evaluation and splicing. In this paper, we omit the specifics of the underlying programming language and we focus on the live coding mechanism. However, we assume that the language is equipped with an evaluation reduction $e \rightsquigarrow p$ that reduces a closed expression e into a value p.

For delayed previews, we construct a delayed expression using splicing. For example, assuming we have a delayed previews $[\![e_0]\!]_x$ and $[\![e_1]\!]_y$. If we need to invoke a member m on e_0 using e_1 as an argument, we construct a new delayed preview $[\![e_0.m(e_1)]\!]_{x,y}$. This operation is akin to expression splicing from meta-programming [X,Y] and can be more formally captured by Contextual Modal Type Theory (CMTT) as outlined below.

Evaluation of previews. The evaluation of previews is defined in Figure 5. Given a dependency graph (V, E), we define a relation $v \downarrow d$ that evaluates a sub-expression corresponding to the node v to a (possibly delayed) preview d.

The auxiliary relation $v \downarrow_{\text{lift}} d$ always evaluates to a delayed preview. If the ordinary evaluation returns a delayed

Figure 5. Rules that define evaluation of previews over a dependency graph for an expression

preview, so does the auxiliary relation (*lift-expr*). If the ordinary evaluation returns a value, the value is wrapped into a delayed preview requiring no variables (*lift-prev*).

Graph node representing a value is evaluated to a value (*val*) and a graph node representing an unbound variable is reduced to a delayed preview that requires the variable and returns its value (*var*).

For member access, we distinguish two cases. If all arguments evaluate to values (*member-val*), then we use the evaluation relation \leadsto , immediately evaluate the member access and produce a value. If one of the arguments is delayed (*member-expr*), because the member access is in the body of a lambda function, then we produce a delayed member access experession that requires the union of the variables required by the individual arguments.

Finally, the evaluation of function values is similar, but requires three cases. If the body can be reduced to a value with no unbound variables (*fun-val*), we return a lambda function that returns the value. If the body requires only the bound variable (*fun-bind*), we return a lambda function with the delayed preview as the body. If the body requires further variables, the result is a delayed preview containing a function.

Caching previews. For simplicity, the relation \downarrow in Figure 5 does not specify how previews are cached and linked to graph nodes. In practice, this is done by maintaining a lookup table $v \to p$ from graph nodes v to (possibly delayed) previews p. Whenever the \downarrow relation is used to obtain a preview for a graph node, we first attempt to find an already evaluated preview using the lookup table. If the preview has not been previously evaluated, we evaluate it and add it to the lookup table.

This means that evaluated previews can be reused in two ways. First, multiple nodes can depend on one sub-graph in a single dependency graph (e.g. if the same sub-expression appears twice in the program). Second, the keys of the lookup table are graph nodes and nodes are reused when a new dependency graph is constructed after the user edits the source code.

Semantics of delayed previews. The focus of this paper is on the design and implementation of a live coding environment, but it is worth noting that the structure of delayed previews is closely linked to the work on Contextual Modal Type Theory (CMTT) [X] and comonads [Y].

In CMTT, $[\Psi]A$ denotes that a proposition A is valid in context Ψ , which is closely related to our delayed preivews written as $[\![A]\!]_{\Psi}$. CMTT defines rules for composing context-dependent propositions that would allow us to express the splicing operation used in (mem-expr). In categorical terms, the context-dependent proposition can be modelled as an indexed comonad [X]. The evaluation of a preview with no context dependencies (built implicitly into our evaluation rules) corresponds to the counit operation of a comonad and would be explicitly written as $[\![A]\!]_{\emptyset} \to A$.

5 Properties

Design and implementaiton of a live coding environment

Something about how dependency graph bits are reused correctly, i.e. if there is a source code change that can change the result, the graph will change.

Something about various code edit operations that will not cause recomputation.

References