#### cite

## Chapter 6 (draft): Notes, Future work, Conclusion

Mark Lemay

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Introduce this section

## Part I

# Symbolic Execution

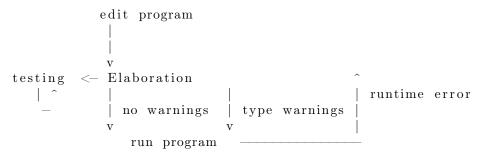
One of the advantages of type checking is the immediacy of feedback. We have outlined here a system that will give warning messages immediately, but requires evaluation to give the detailed error messages that are most helpful when correcting a program. This is especially important if the user wants to use the system as a proof language, and will not generally execute their proofs. A symbolic evaluation system recaptures some of that quicker feed back, by having a system that passively tries to find errors. This ideal workflow appears in table 1.

Since this procedure operates over the cast language, we must decide what constitutes a reasonable testing environment

A one hole context can be defined for the cast languages presented in this thesis C[-] and we can say that an error is observable if  $\vdash a : A$  and  $\vdash C[a] : B$  and  $C[a] \leadsto_* b$  and **Blame**  $\ell \circ b$  for some  $\ell \in lables(a)$  but  $\ell \notin lables(C[-])$ .

This process is semi-decidable in general (by enumerating all well typed syntax). But testing every term is infeasibly inefficient for functions and types. An approximate approach can build partial contexts, that are correct up to some constraint. For instance,

- If we have the "empty" type,  $\lambda x \Rightarrow \star ::_{\ell} x : (x : \star) \to x$ , then we can observe an error by applying x of type  $x : \star$  to the term and  $x = \to -$ , which would correspond to the context  $[\lambda x \Rightarrow \star ::_{\ell} x] (\star \to \star) \leadsto_{\ast} \star ::_{\star,\ell} (\star \to \star)$  and **Blame**  $\ell . \star ::_{\star,\ell} (\star \to \star)$
- This reasoning can be extended to higher order functions  $\lambda f \Rightarrow \star ::_{\ell} f \star :: \star : (\star \to \star) \to \star$ , then we can observe an error by applying f of type  $f : \star \to \star$  to the term and  $f \star = \to -$ , which would correspond to the context  $[\lambda f \Rightarrow \star ::_{\ell} f \star :: \star] (\lambda x \Rightarrow (\star \to \star)) \leadsto_{\star} \star ::_{\star,\ell} (\star \to \star)$  and **Blame**  $\ell . \star ::_{\star,\ell} (\star \to \star)$
- Observing a higher order input directly,  $\lambda f \Rightarrow f(!!) : (\star \to \star) \to \star$ , and can observe an error by inspecting



better graphics

Figure 1: Ideal Workflow

its input, which would correspond to the context  $[\lambda f \Rightarrow f(!!)](\lambda x \Rightarrow x) \rightsquigarrow_*!!$  which is blamable. Where !! will stand infor any blamabme term (here !! =  $\star$  ::  $(\star \to \star)$  ::  $\star$ )

- Similarly with dependent types,  $!! \to \star$  :  $\star$ , can observe an error by inspecting its input, which would correspond to the context  $((\lambda x \Rightarrow x :: \star) :: [!! \to \star] :: (\star \to \star)) \star \leadsto_{\star} \star :: [!!] :: \star$  which is blamable.
- Similarly we will allow extraction from the dependent body of a function type  $(b : \mathbb{B}_c) \to b \star !! \star : \star$ , by symbolically applying b where  $b : \mathbb{B}_c, b \star = y$  leaving  $y !! \star$  wich can obseve blame via y's first argument.
- data can be observed incrementally, and paths along data will confirm that the data conductors are consistent

The procedure insists that the following constraints hold,

- variables and assignments are type correct
- observable different outputs must come from observably different inputs (in the case of dependent function types, the argument should be considered as an input)

•

This seems to be good because the procedure,

- can guide toward the labels of interest, for instance we can move to labels that have not yet observed a concrete error. Terms without lables can be skipped entirely.
- can choose assignments strategically avoiding or activating blame as desired
- Since examples are built up partially the partial contexts can avoid introducing their own blame by construction
- handle higher order functions, recursions, and self reference gracefully. For instance  $f: Nat \to Nat$ , f(f 0) = 1 and f(3) = 3 if there is an assignment that implies  $f 0 \neq 3$

However the procedure is unsound, it will flag errors that are not possible to realize in a context,

• However since there is no way for a terms within the cast language to "observe" a distinction between the type formers plausible environments cannot always be realized back to a term that would witness the bad behavior. For instance, the environment  $F: \star \to \star, F \star = \star \to \star, F \ (\star \to \star) = \star$ , which might be needed to explore the term with casts like  $\lambda F \Rightarrow \dots :: \star :: F \ (\star \to \star) \dots :: (\star \to \star) :: F \star : (\star \to \star) \to \dots$ , cannot be realized as a closed term. In this way the environment is stronger then the cast language. The environment reflects a term language that has a type case construct.

#### add non termination as a source of unrealizability

• Additionally, working symbolically can move past evaluations that would block blame in a context. For instance that procedure outlined above would allow !! to be reached in  $(\lambda xf \Rightarrow f(!!)) loop : \star \to \star$  even though there is no context that would allow this to cause blame.

#### add por as a source of unrealizability

• Subtly a version of parallel-or can be generated via assignment even though such a term is unconstructable in the language,  $por : \mathbb{B} \to \mathbb{B} \to \mathbb{B}$ ,  $por\ loop\ tt = tt$ ,  $por\ tt\ loop\ = tt$ ,  $por\ ff\ ff\ = ff$ . Here all assignments are well typed, and each output can be differentiated by a different input.

## 1 Related and Future Work

While formalizing a complete and efficient testing procedure along these lines is still future work. There are likely insights to be gained from the large body of research on symbolic execution, especially work that deals with typed higher order functions. A fully formal account would deal with a formal semantics of the cast language.

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#### 1.1 Testing

Many of the testing strategies for typed functional programming trace their heritage to **property based** testing in QuickCheck [CH01]. Property based testing involves writing functions that encode the properties of interest, and then randomly testing those functions.

- QuickChick <sup>1</sup> [DHL<sup>+</sup>14][LPP17, LGWH<sup>+</sup>17, Lam18] uses type-level predicates to construct generators with soundness and completeness properties, but without support for higher order functions. However, testing requires building types classes that establish the properties needed by the testing framework such as decidable equality. This is presumably out of reach of novice Coq users.
  - Current work in this area uses coverage guided techniques in [LHP19] like those in symbolic execution.
     More recently Benjamin Pierce has used Afl on compiled Coq code as a way to generate counter examples<sup>2</sup>.
- [DHT03] added QuickCheck style testing to Agda 1.

# review the there's a now?

#### 1.2 Symbolic Execution

Symbolic evaluation is a technique to efficiently extract errors from programs. Usually this happens in the context of an imperative language with the assistance of an SMT solver. Symbolic evaluation can be supplemented with other techniques and a rich literature exists on the topic.

The situation described in this chapter is unusual from the perspective of symbolic execution:

- the number of blamable source positions is limited by the location tags. Thus the search is error guided, rather then coverage guided.
- The language is dependently typed. Often the language under test is untyped.
- The language needs higher order execution. often the research in this area focuses on base types that are efficiently handleable with an SMT solver.

This limits the prior work to relatively few papers

- A Symbolic execution engine for Haskell is presented in [HXB<sup>+</sup>19], but at the time of publication it did not support higher order functions.
- A system for handling higher order functions is presented in [NTHVH17], however the system is designed for Racket and is untyped. Additionally it seems that there might be a state space explosion in the presence of higher order functions.
- [?] extended and corrected some issues with [NTHVH17], but still works in a untyped environment. The authors note that there is still a lot of room to improve performance.
- Closest to the goal here, [LT20] uses game semantics to build a symbolic execution engine for a subset of ML with some nice theoretical properties.
- An version of the above procedure was experimented with in the extended abstract , however conjectures made in that preliminary work were false (the procedure was unsound). The underlying languages has improved substantially since that work.

The subtle unsoundness hints that the approach presented here could be revised in terms of games semantics (perhaps along lines like [LT20]). Though game semantics for dependent types is a complicated subject in and of itself. Additionally is seems helpful to allow programmers to program their own solvers that could greatly increase the search speed.

cite

<sup>&</sup>lt;sup>1</sup>https://github.com/QuickChick/QuickChick

<sup>&</sup>lt;sup>2</sup>https://www.youtube.com/watch?v=dfZ94N0hS4I

## Part II

## Runtime Poof Search

Just as "obvious" equalities are missing from the definitional relation, "obvious" proofs and programs are not always conveniently available to the programmer. For instance, in Agda it is possible to write a sorting function quickly using simple types. With effort is it possible to prove that sorting procedure correct by rewriting it with the necessarily dependently typed invariants. However, very little is offered in between. The problem is magnified if module boundaries hide the implementation details of a function, since those details are exactly what is needed to make a proof! This is especially important for larger scale software where a library may require proof terms that while true are not provable from the exports of other libraries.

The solution proposed here is additional syntax that will search for a term of the type when resolved at runtime. Given the sorting function

```
\mathtt{sort}:\mathtt{List}\,\mathbb{N}\to\mathtt{List}\,\mathbb{N}
```

and given the first order predicate that

```
{\tt IsSorted}: {\tt List}\, \mathbb{N} \to *
```

then it is possible to assert that sort behaves as expected with

$$\lambda x.$$
? :  $(x : \texttt{List} \, \mathbb{N}) \to \texttt{IsSorted} \, (\texttt{sort} x)$ 

This term will act like any other function at runtime, given a List input the function will verify that the sort correctly handled that input, or the term will give an error, or non-terminate.

Additionally, this would allow simple prototyping form first order specification. For instance,

```
\begin{aligned} & \textit{data} \; \texttt{Mult} : \mathbb{N} \to \mathbb{N} \to \mathbb{N} \to * \; where \\ & \texttt{base} : (x : \mathbb{N}) \to \texttt{Mult} \; 0 \; x \; 0 \\ & \texttt{suc} : (x \, y \, z : \mathbb{N}) \to \texttt{Mult} \, x \, y \, z \to \texttt{Mult} \; (1 + x) \; y \, (y + z) \end{aligned}
```

can be used to prototype

$$\mathtt{div} = \lambda z. \lambda x. \mathtt{fst} \left( ? : \sum y : \mathbb{N}. \mathtt{Mult} x \, y \, z \right)$$

The symbolic execution described above can precompute many of these solutions in advance. In some cases it is possible to find and report a contradiction.

Experiments along these lines have been limited to ground data types, and a an arbitrary solution is fixed for every type. Ground data types do not need to to worry about the path equalities since all the constructors will be concrete.

Non ground data can be very hard to work with when functions or function types are considered. For instance,

$$?: \sum f: \mathbb{N} \rightarrow \mathbb{N}. \mathrm{Id}\left(f, \lambda x. x + 1\right) \& \mathrm{Id}\left(f, \lambda x. 1 + x\right)$$

It is tempting to make the? operator sensitive to more then just the type. For instance,

```
n : Nat;
n = ?;

pr : Id Nat n 1;
pr = refl Nat 1;
```

will likely give the waning error "n = ?= 1 in Id Nat  $\underline{n}$  1". It will then likely give the runtime error "0 = != 1". Since the only information to solve ? is the type Nat and an arbitrary term of type nat will be 0. Most users would expect n to be solved for 1.

However constraints assigned in this manner can be extremely non-local. For instance,

```
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```

```
\begin{array}{l} n \ : \ Nat; \\ n \ = \ ?; \\ \\ \dots \\ pr \ : \ Id \ Nat \ n \ 1; \\ pr \ = \ refl \ Nat \ 1; \\ \dots \\ pr2 \ : \ Id \ Nat \ n \ 2; \\ pr2 \ = \ refl \ Nat \ 2; \\ \\ And \ thing \ become \ even \ more \ complicated \ when \ solving \ is \ interleaved \ with \ computation. \ For \ instance, \\ n \ : \ Nat; \\ n \ = \ ?; \\ prf \ : \ Nat \ -> \ Nat \ ; \\ prf \ x \ = \ (\ \ \_ \Rightarrow x) \ (\ refl \ Nat \ x \ : \ Id \ Nat \ n \ x); \end{array}
```

### 2 Prior Work

Proof search is often used for static term generation in dependently typed languages (for instance Coq tactics). A first order theorem prover is attached to Agda in [Nor07]. However it is rare to make those features available at runtime

Logic programing languages such as Prolog<sup>3</sup>, Datalog<sup>4</sup>, and miniKanren<sup>5</sup> use "proof search" as their primary method of computation. Dependent data types can be seen as a kind of logical programming predicate. The Twelf project<sup>6</sup> makes use of runtime proof search and has some support for dependent types, but the underling theory cannot be considered full-spectrum. The only full spectrum systems that preport to handle solving in this way are the gradual dependent type work, though it is unclear how that work handles the non locality of constraints given their local? operator.

## Part III

## Convenience

implicit function arguments

### Part IV

## Future work

### 3 Effects

The last and biggest hurdle to bring dependent types into a mainstream programming language is by providing a reasonable way to handle effects. Though dependent types and effects have been studied I am not aware of any full spectrum system that has implemented those systems. It is not even completely clear how to best to add an effect system into Haskell, the closest "mainstream" language to the one studied here.

 $<sup>^3 \</sup>mathrm{https://www.swi-prolog.org/}$ 

 $<sup>^4</sup> https://docs.racket-lang.org/datalog/$ 

<sup>&</sup>lt;sup>5</sup>http://minikanren.org/

 $<sup>^6</sup>$ http://twelf.org/wiki/Main\_Page

dox from s should d speciWhile trying to carefully to avoid effects in this thesis, we still have encountered 2 important effects, non-termination and blame based error.

Non-termination is treated allowed, but it would be better to have it work in the same framework as equational constraints, namely warn when non-termination is possible, and try to find slow running code via symbolic execution. Then we could say without caveat "programs without warnings are proofs".

Blame based errors aren't handled, handling effects in a dependent type system is not strait forward since the handling construct can observe differences that should not otherwise be apparent.

One of the difficulties of an effect system for dependent types is expressing the definitional equalities of the effect modality. Is print "hello";print "world" = print "helloworld": Writer Unit? by differing checks to runtime these issues can be avoided. further effects risk making computation mathematically inelegant, In this thesis we avoided this inelegance with an additional cast syntax and blame relation, This strategy could perhaps be applied to more interesting effect systems.

Both the symbolic execution and search above could be considered in terms of an effect in an effect system. proof search specifically could be localized though an effect modality.

## 4 Semantics

#### References

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# $\begin{array}{c} {\rm Part} \ {\rm V} \\ {\rm TODO} \end{array}$

## Todo list

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require a minimum output	
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add non termination as a source of unrealizability .	
add por as a source of unrealizability	
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## Part VI

# notes

## Part VII

# unused

More subtle is that the procedure described here will allow f to observe parallel or, even though parallel or cannot be constructed within the language. This hints that the approach presented here could be revised in terms of games semantics (perhaps along lines like [LT20]). Though game semantics for dependent types is a complicated subject in and of itself.

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