

Give an example of a C program that uses type coercion

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- P3 deadline tonight
- P4 released "Monday morning", i.e. Sunday @ 11:59 PM + 1 minute

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YHP

Type Analysis

CONSTRUCTION

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Discuss Type Systems

- What they are
- Why we use them

Type Specification (optional)

• How we communicate type systems

You Should Know

- What a type system is
- How type systems effect semantics

Enforcing Type Systems

• Design points

Type Analysis

- Type checking
- Type inference / synthesis

Enforcing Type Systems Type Analysis

Language property: how much enforcement / checking to do?

- Idea 1: check what you can, allow uncertainty
- Idea 2: check what you can, disallow uncertainty completely
- Idea 3: check what you can, force user to dispel uncertainty

e.g. Haskell

e.g. Java, Rust

Some languages allow an explicit means to "escape" the type system

• Typecasting – allow one type to be used as another type

Cross-casting (static check in Java)

 $\sum_{i=1}^{n}$ Apple dApp = (Apple) f; Runtime check

Cross-casting (static check in Java)

Apple $a = new Apple()$; \int Orange $o =$ (Orange) a; Compiler check **Downcasting (dynamic check in Java)** Fruit $f = new Apple()$; if (rand()) { $f = new Orange()$; \leftarrow } Δ Apple dApp = (Apple) f; Runtime check

Strongly-Typed vs Weakly-Typed Enforcing Types

Colloquial classification of a language's type system

- Degree to which type errors are allowed to happen at runtime
- Continuum without precise definitions

- Has a precise definition
	- All successful operations must be allowed by the type system
- Java was explicitly designed to be type safe
	- A variable of some type can only be used as that type without causing an error
- C is very much not type safe
- C++ isn't either but it is safer

Type Safety Violations Type Enforcement

C Format specifier printf("%s", 1);

Memory safety

struct big{

int a[1000000];

};

```
struct big * b = malloc(1);
```
C++

}

Unchecked casts class T1 { char a }; class T2 { int b }; int main { $T1 * myT1 = new T1();$ T2 $*$ myT2 = new T2(); $myT1 = (T1*) myT2;$

Research on Types Type Checking

• Some CS Deparments have a "PLT" focus: "Programming Languages and Types"

Abstract

We present Logically Qualified Data Types, abbreviated to Liquid Types, a system that combines Hindley-Milner type inference with Predicate Abstraction to automatically infer dependent types precise enough to prove a variety of safety properties. Liquid types allow programmers to reap many of the benefits of dependent types, namely static verification of critical properties and the elimination of expensive run-time checks, without the heavy price of manual annotation. We have implemented liquid type inference in DSOLVE, which takes as input an OCAML program and a set of logical qualifiers and infers dependent types for the expressions in the OCAML program. To demonstrate the utility of our approach, we describe experiments using DSOLVE to statically verify the safety of array accesses on a set of OCAML benchmarks that were previously annotated with dependent types as part of the DML project. We show that when used in conjunction with a fixed set of array bounds checking qualifiers. DSQLVE reduces the amount of manual annotation required for proving safety from 31% of program text to under 1%

Categories and Subject Descriptors D.2.4 [Software Engineeringl: Software/Program Verification: F.3.1 [Logics and Meanings] of Programs]: Specifying and Verifying and Reasoning about Pro-

General Terms Languages, Reliability, Verification

Keywords Dependent Types, Hindley-Milner, Predicate Abstraction, Type Inference

1. Introduction

Modern functional programming languages, like ML and Haskell, have many features that dramatically improve programmer productivity and software reliability. Two of the most significant are strong static typing, which detects a host of errors at compile-time, and type inference, which (almost) eliminates the burden of annotating the program with type information, thus delivering the benefits of strong static typing for free.

^{*}This work was supported by NSF CAREER grant CCF-0644361, NSF PDOS grant CNS-0720802, NSF Collaborative grant CCF-0702603, and a gift from Microsoft Research.

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The utility of these type systems stems from their ability to predict, at compile-time, invariants about the run-time values com puted by the program. Unfortunately, classical type systems only capture relatively coarse invariants. For example, the system can express the fact that a variable i is of the type int, meaning that it is always an integer, but not that it is always an integer within a certain range, say between 1 and 99. Thus, the type system is unable to statically ensure the safety of critical operations, such as a division by i, or the accessing of an array a of size 100 at an index i. Instead, the language can only provide a weaker dynamic safety guarantee at the additional cost of high performance overhead. In an exciting development, several authors have proposed the use of *dependent types* [20] as a mechanism for enhancing the expressivity of type systems [14, 27, 2, 22, 10]. Such a system can express the fact

i :: $\{\nu : \text{int} \mid 1 \leq \nu \wedge \nu \leq 99\}$

which is the usual type int together with a refinement stating that the run-time value of i is an always an integer between 1 and 99. Pfenning and Xi devised DML, a practical way to integrate such types into ML, and demonstrated that they could be used to recover static guarantees about the safety of array accesses, while simultaneously making the program significantly faster by eliminating run-time checking overhead [27]. However, these benefits came at the price of automatic inference. In the DML benchmarks, about 31% of the code (or 17% by number of lines) is manual annotations that the typechecker needs to prove safety. We believe that this nontrivial annotation burden has hampered the adoption of dependent types despite their safety and performance benefits.

We present Logically Qualified Data Types, abbreviated to Liquid Types, a system for automatically inferring dependent types pro cise enough to prove a variety of safety properties, thereby allowing programmers to reap many of the benefits of dependent types without paying the heavy price of manual annotation. The heart of our inference algorithm is a technique for blending Hindley-Milner type inference with *predicate abstraction*, a technique for synthesizing loop invariants for imperative programs that forms the algorithmic core of several software model checkers [3, 16, 4, 29, 17]. Our system takes as input a program and a set of logical qualifiers which are simple boolean predicates over the program variables, a special value variable ν , and a special placeholder variable \star that can be instantiated with program variables. The system then infers liquid types, which are dependent types where the refinement predicates are *conjunctions* of the logical qualifiers.

In our system, type checking and inference are decidable for three reasons (Section 3). First, we use a conservative but decidable notion of subtyping, where we reduce the subtyping of arbitrary dependent types to a set of implication checks over base types, each of which is deemed to hold if and only if an embedding of the implication into a decidable logic vields a valid formula in the logic. Second, an expression has a valid liquid type derivation only if it has a valid ML type derivation, and the dependent type

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• A type enhanced with a predicate which must hold for any element of that type

$$
f: \mathbb{N} \to \{n: \mathbb{N} \mid n\%2 = 0\}
$$

- Could imagine enhancing a type system with annotations for all kinds of properties
	- Single-use variable
	- High security/low security (non-interference)

More Research on Types Type Checking

Piggybacking on Type Checking Type Checking

- Type checking is a good place to get extra programmer hints:
	- Programmers are already familiar with typing logic
	- The analysis is already well-formulated

Formal Type Systems End Detour: Done with Ungraded Material

Generate appropriate code for operations

- $A + B$
	- String concatenation? Integer addition? Floating-point addition

Catch runtime errors / security

- Make sure operations are sensible
- Augment type system with addition checks

Type Analysis

- Assigning types to expressions
- Flavors:
	- Type synthesis get type of an AST node from it's children
	- Type inference get type of an AST node from it's use context

Type Checking

• Ensure that type of a construct is allowed by the type system

Implementing Our Type Checker Type Checking

Implementing Typing Our Type System

Structurally similar to nameAnalysis

- Historically, intermingled with nameAnalysis
- Done as part of AST attribute "decoration"

Add a typeCheck method to AST nodes

- Recursively walk the AST checking subtypes
	- "Inside out" analysis
	- Attach types to nodes
	- Propagate an error symbol

- Get the type of the LHS
- Get the type of the RHS
- Check that the types are compatible for the operator
- Set the *kind* of the node be a value
- Set the *type* of the node to be the type of the operation's result

- Cannot be wrong
	- Just pass the type of the literal up the tree

- Look up the type of the declaration
	- There should be a symbol "linked" to the node
- Pass symbol type up the tree

Function Calls Implementing Type Checking

- Get type of each actual
- Match against formals of the called function's symbol
- Propagate return type to parent node

Statements Implementing Type Checking

Always have void type

- Make sure to check child expression
- No type to propagate
- Some versions of analysis may propagate boolean: error / no error

Other AST Node Types Implementing Type Checking

Follow these same principles

- Ensure that children are well-typed
- Apply a combination rule
	- If valid: infer a type and propagate out
- If invalid: propagate error
 $m q y l e \times m q q N$
 $Q q N$

 w^2
 $\frac{1}{2}$ $\int e^{\frac{1}{2}(\frac{1}{2} - \frac{1}{2})} e^{\frac{1}{2}(\frac{1}{2} - \frac{1}{2})} e^{-\frac{1}{2}(\frac{1}{2} - \frac{1}{2})}$

Exercise: Draw Type Analysis Bonus Exercise

Handling Errors Implementing Type Checking

- We'd like all *distinct* errors at the same time
	- Don't give up at the first error
	- Don't report the same error multiple times
- When you get error as an operand
	- Don't (re)report an error
	- Again, pass **error** up the tree

The difference between…

Neither operand works with the operator

- We'd like all *distinct* errors at the same time
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- When you get error as an operand
	- Don't (re)report an error
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Having explorer two semantic analyses, let's generalize

• What's the limit of semantic analysis, especially error checking?