

Optimizing Message Sends in Object Oriented Languages through Type Invariant Region Analysis ^{*}

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Abstract. Compilers for Object-Oriented Languages optimize dynamic message sends through run-time type testing when they cannot precisely infer the message receiver type. While run-time type testing optimizes dynamic messages, it still results in redundant type testing. We propose a flow-sensitive analysis which calculates regions in a program where an object's type is statically unknown but invariant. The type check is hoisted at the entry point of a region and its results are shared over all dynamic call-sites in the region. We implemented this optimization in the Vortex compiler and demonstrate its effectiveness on five large Cecil benchmarks. We observed an average speed-up of 9% and 18% speed-up in the best case.

1 Introduction and Related Work

Typical programs written in Object-Oriented Programming Languages (OOPs) result in a large number of dynamic message sends due to lack of static type information. Dynamic message sends can be optimized in OOPs by using a compile-time interprocedural *type inference* [24] [22] [21] [23] [3] [12]. However, at best only 60% of the message receiver types can be precisely inferred¹.

When a type inferencer cannot statically bind a message send, the compiler inserts some sort of run-time type test. If the set of possible receiver types for a message send is small, the compiler will inline a series of type tests [2]. Otherwise, the compiler generates a dynamic run-time type test through a (polymorphic) inline cache [14] [19].

In all the approaches above, some sort of run-time type check is required when a dynamic message cannot be precisely inferred. However, neither of these

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¹ DeFouw et al. [13] showed that only 40-60% of the dynamic call-sites in the Cecil [6] OOP could be inferred. Agesen and Hölzle [2] conducted a similar study on the Self [25] OOP and found that 30-40% of all dispatches remained in their tiny benchmarks, 50-60% in their small to medium benchmarks, and 25-60% in their large benchmarks.

techniques take into account type check redundancy when a type of a receiver object does not mutate (or change) within a large program region. The fact is, not all method calls are dynamically bound during run-time. Grove et al. [18] showed that 50% of the dynamic dispatches in their Cecil programs had a single receiver class. This would indicate that many dynamically typed objects, although statically unknown in their type binding, have a fixed type during a particular region in the program. In these cases, type tests are only needed at the first use of the object within the region. Run-time type tests or cache probes for subsequent uses of these objects are extraneous.

A partial redundancy elimination (PRE) of redundant type tests can be performed. One such optimization is *splitting* [10]. While splitting can eliminate redundant type tests, code between the first test and redundant test must be duplicated. If the benefiting code is far away, the increase in code size may outweigh the benefits of splitting. This would indicate the need for a new technique based on PRE that can be applied to large regions of code. Our framework, presented in the next section, is motivated by this fact.

2 Type Invariant Region Analysis

In order to eliminate unneeded type tests, one could perform a static analysis to determine the program region where the type (although statically unknown) is invariant once it is dynamically fixed. Methods dispatched on objects whose types are fixed in a particular program region could be directly called; saving on the overhead of type tests and/or dynamic dispatches. To maximize the call-sites that are covered by the type invariant region, one could take the transitive closure of the object's reaching definitions and use a flow-sensitive analysis for better precision. After calculating this information, we can perform one type test and share its results for all call-sites in the region². The end result is fewer type checks and more direct calls in a given program. Because we generate more direct calls, the performance greatly improves. The following motivating example shows the benefit of our approach.

2.1 Motivating Example

Figure 1 shows part of a list library written in Cecil. The method *reverse* implements a recursive list reversal on a list with data polymorphism³ (the integer implementation is shown). Because the second parameter (*Alist*) has data polymorphism, the compiler cannot statically bind the tail recursive call.

One observation that can be made from this example is that the types of the formal parameters are unknown at compile-time, but known at the first invocation of the method. In other words, they do not mutate across recursive invocations. Most OO languages with functional flavor (such as Cecil) support

² A more formal definition of region is given later.

³ The *mList* object can store a heterogeneous list of integers, floats, strings, etc.

this model since formal parameters are immutable [6]. This allows us to statically define a region on the *Alist* formal parameter as illustrated in figure 2. Next, we can fix the targets of each call-site in the region at the first recursive invocation of reverse. Because the type of Alist is immutable, we are guaranteed that the targets for the Alist region will not change across recursive invocations. This allows us to cache the method targets and save on the method lookup overhead across recursive invocations leading to better run-time performance. We also observe that in many typical applications, the parameter types do not always mutate across *multiple* invocations of the same method at *different* call-sites.

```

method reverse(front@:int, Alist@:m_list[int]) : m_list[int] {
  if(Alist.is_empty, { ^Alist; });
  let var Blist := reverse(first(Alist),rest(Alist));
  Blist.add_first(front);
  Blist;
}

```

Fig. 1. Motivating example Cecil code

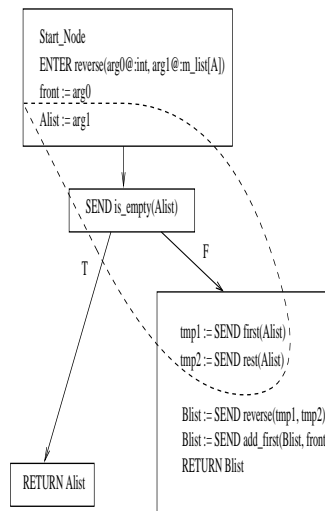


Fig. 2. Motivating Example Control Flow Graph with Corresponding Type Invariant Region for Parameter Alist.

2.2 Terms and Definitions

- A **control flow graph** (CFG) is a directed graph $G = (N, E, START, END)$, where N is the set of nodes, E is the set of flow edges, $START \in N$, and $END \in N$.
- A **definition** of an object instance x is a statement that assigns, or may assign, a value (as well as type) to x .
- When we make an assignment to an object instance x , the set of *definitions* that *may* determine the *type* and *value* of x are called the set of **reaching definitions** for x .
- The **receiver set** of a message send is the set of object definitions that determine the destination of the message. This set consists of 1-tuples for **singly-dispatched** message sends and n-tuples of object definitions for **multiple-dispatched** (or multi-method) [8] message sends.
- The program **call graph** is a CFG that represents the (inter-procedural) calling relationships between the program’s procedures.
- The **type binding** (or simply type) of any object used in the receiver set for a singly- or multiple-dispatched message send is determined solely by a data-flow (e.g. an initialization, assignment, side-effect through a message send, etc.) [4].

2.3 Formal Framework

Our goal is to detect the portion of a method’s CFG in which the type of a call-site’s dispatching argument is invariant. This portion of the CFG is referred to as the *type invariant region*. After detecting the type invariant region of an object, run-time type tests can be hoisted to the top of the region and shared with all call-sites that are dispatched on the type of the object.

For purposes of efficient code hoisting, we require a flow-sensitive analysis. We are only interested in dynamic call-sites whose types *must* remain invariant based on the type of the formal parameter, not those that *may* remain invariant (computed by a flow-insensitive analysis).

Definition 1. $region(n) = \{A \text{ Method call-site } p \text{ such that the callee of } p \text{ can be uniquely inferred knowing the type binding of formal parameter } n\}$

Consider the method call-site $p(a, b, c)$ within an outer call $q(n)$ in figure 3. Assume $p(a,b,c)$ is dispatched on arguments a, b and c . Our goal is to determine if the call-site p belongs to the $region(n)$. If the types of a, b and c can be uniquely inferred knowing the type binding of n , then we can deduce the callee of method p and avoid a dynamic dispatch. We now propose a property that must be satisfied in order to determine the region of a formal parameter.

Lemma 1. *Refer to figure 3. Let n^i denote a possible type binding of formal parameter n ; a_j denote a reaching definition of a ; b_k denote a reaching definition of b ; and c_r denote a reaching definition of c to a call-site $p(a,b,c)$.*

A call-site p belongs to the $region(n)$ only if $\forall a_j, b_k, c_r, type(n^i) \Rightarrow type(a_j)$ and $type(n^i) \Rightarrow type(b_k)$ and $type(n^i) \Rightarrow type(c_r)$ such that the inferred type bindings for a, b and c lead to a single callee of p .

```

method q(n) : void {
  ...
  p(a, b, c);
  ...
}

```

Fig. 3. Formal Framework Motivating Example

Proof:

It is obvious that if we can infer the types of a , b , and c in $p(a, b, c)$ by using a type binding of n that leads to a single callee for p , then we can uniquely bind the dispatch of p . In order for this to be true, all the reaching definitions of parameters a , b and c must be type deducible for a given type binding n_i of n and must lead to a single callee for p .

Definition 2. $\text{typededucible}(n^i) = \{ \text{The set of all definitions } d_i \text{ such that the type of } d_i \text{ can be uniquely inferred using a given type binding } n^i \}$

Corollary 1. $p(a, b, c)$ is in the region of n if for a given type binding n^i of n , $\forall a_j, b_k$ and $c_r \in \text{typededucible}(n^i)$ leading to a single callee for p where a_j, b_k and c_r represent reaching definitions of a, b and c respectively at call-site p .

In its full generality, computing $\text{typededucible}(n^i)$ and $\text{region}(n^i)$ poses the problem of *flow-sensitive* type inferencing using reaching definitions. Several algorithms for inferring types and constructing call graphs for OOPs have been proposed, however they are *flow-insensitive* approaches [4] [3]. *Flow-sensitive* analysis is needed to avoid a loss of precision in our framework. In order to build a flow-sensitive call graph, reaching definitions need to be calculated. However, solving such a problem in the interprocedural scope can have a considerable complexity as shown by Agesen⁴. Moreover, there is no empirical data yet that demonstrates the tractability of flow-sensitive call graph construction using reaching definitions [1]. Thus, we limit our analysis to the intraprocedural scope. Also we propose a framework based on reaching definitions of values rather than types. This allows us to avoid type inferencing. Although this approach is certainly less precise, we contend that in the intraprocedural sense the loss of precision is quite tolerable due to the limited scopes and type locality of the values involved.

In our value based approach, we use the closure of reaching definitions of formal parameter n . We introduce a concept of $\text{reaching_defs_copy}(n)$.

Definition 3. $\text{reaching_defs_copy}(n)$ is the set of all values that are copies of reaching definitions of n within the callers' scope of method $q(n)$.

⁴ Reaching definitions and the call graph must be built simultaneously [1]. Unfortunately this adds a layer of complexity to the already complex iterative refinement that resolves the type and call graph mutually dependent problems [17] [16]. Agesen also warns that it is easy to find code examples with inferior asymptotic precision [1].

Thus, $k \in \text{reaching_defs_copy}(n)$ if and only if either there exists an assignment $k = n$ or $k = r$ where, $r \in \text{reaching_defs_copy}(n)$.

Using these definitions, we can restate the region of n as follows:

Corollary 2. $p(a, b, c) \in \text{region}(n)$ if $\forall a_j, b_k$ and $c_r \in \text{reaching_defs_copy}(n)$ leading to a single callee for p where a_j, b_k and c_r represent reaching definitions of a, b and c respectively at call-site p .

Proof:

It is obvious that the set $\text{reaching_defs_copy}(n) \subseteq \bigcup_i \text{typededucible}(n^i)$, thus satisfies corollary 1.

In order to effectively use corollary 2, we must be able to include all call-sites that could be inferred using the transitive closure of the set $\text{reaching_defs_copy}(n)$. Finally, if the outer call-site $q(n)$ had multiple dispatching arguments (such as in $q(n, m)$) corollary 2 still holds, except that we now have a definition of the region in terms of a cartesian product of type bindings n and m . The above definitions could be trivially extended to these cases.

Lemma 2. *The type for an object definition x is said to be invariant from a program point y to a program point z if and only if there does not exist a path from y to z that kills definition x .*

Proof:

When considering objects that are used in singly- or multiple-dispatched message sends, the type for an object is determined solely on a data-flow (e.g., an initialization, assignment, side-effect through a message send, etc.) [4]. In order for x to mutate, there must exist a path from y to z , such that definition x gets killed. Conversely, if there does not exist a redefinition of x along any path between y and z , then the type for x must remain invariant between y and z if and only if type is determined solely by a data-flow.

Definition 4. *Using the definitions of x, y , and z from Lemma 2, we describe the type invariant region for x as all the paths from y to z if and only if lemma 2 holds.*

The above lemma and definition for *type invariant region* is applicable for singly- or multiple-dispatched message sends. *Predicate dispatching* [7] [15], another (although less common) dispatching technique, is not considered here ⁵.

2.4 Algorithms for Type Invariant Region Detection

The implementation algorithms are presented in figures 4 and 5. The first algorithm takes as input $\text{DefUse}(X)$ which is calculated using a standard du chain algorithm [5]. Next, we calculate $\text{region_reaching_defs_copy}(\text{region_id})$ in figure 4. In the first loop we are seeding $\text{region_reaching_defs_copy}(n)$ with all

⁵ Determining the type-invariant region for an object used in predicate dispatching is undecidable without loss of generality. A proof was omitted due to space limitations. The interested reader may request the proof from the authors.

immutable objects⁶. In the second loop we calculate their transitive closure. After `region_reaching_defs_copy(region_id)` is calculated, we can calculate `region(n)` (definition 1) in figure 5. We use a table called `region_table` that maps a receiver set to a set of message sends. We initially seed `region_table` with the range of `region_reaching_defs_copy`. Next, we traverse a CFG and examine each dynamic message send. Whenever we encounter a dynamic message send in which its receiver set is in the domain of `region_table`, we add it to the appropriate message set. After traversing the CFG, `region(n)` is defined.

Algorithm 1: Compute_Region_Reaching_defs_copy(region_cnt)

Input: $DefUse(X)$,
 $DefUse(X)$ is a table that maps the object definition X to a set of reaching object definitions used by X
Output: `region_reaching_defs_copy(region_cnt)`,
table `region_reaching_defs_copy(region_id)` maps $region_id \in integer \rightarrow reaching_defs_copy(n)$,
`region_cnt` = the number of regions detected in a CFG and
table `region_reaching_defs_copy` maps $(1 \leq region_id \leq region_cnt) \rightarrow reaching_defs_copy(n)$.

```

change := true;
region_cnt := 1;

for each  $Z \in Domain(DefUse(X))$  do
    if(  $Z$  is an immutable object ) then
        region_reaching_defs_copy(region_cnt) := region_reaching_defs_copy(region_cnt)  $\cup Z$ ;
        increment region_cnt;
    end if
end do

while( change = true ) do
    change := false;
    for each  $Z \in Domain(DefUse(X))$  do
        let  $Y := DefUse(Z)$ ;
        if(  $\neg(Z \in Y)$  ) then
            for each  $i \in region\_reaching\_defs\_copy(n)$  do
                if(  $Y \subseteq region\_reaching\_defs\_copy(i)$  ) then
                    region_reaching_defs_copy(i) := region_reaching_defs_copy(i)  $\cup Z$ ;
                    change := true;
                end if
            end do
        end if
    end do
end do
end do

```

Fig. 4. Algorithm 1: Compute `region_reaching_defs_copy(region_cnt)`

⁶ In Cecil these include all formal parameters, user-specified immutables, and compiler-generated temporaries. All other objects (e.g., variables) in Cecil are mutable [6].

Algorithm 2: ComputeRegions

Input: The control flow graph (CFG), $region_reaching_defs_copy(region_cnt)$ (computed in Algorithm 1).

Output: $region(n)$ (definition 1).

let $message_set$ be a set of nodes $N \in CFG$ such that N is a dynamic message send;
let $region_table$ be a table that maps $receiver_set \rightarrow message_set$;

```

for each  $i \in region\_reaching\_defs\_copy(n)$  do
  Domain( $region\_table$ ) := Domain( $region\_table$ )  $\cup$   $reaching\_defs\_copy(i)$ 

for each node  $N \in message\_set$  do begin
  let  $R := receiver\_set(N)$ ;
  if(  $R \in Domain(region\_table)$ ) then
     $region\_table(R) := region\_table(R) \cup N$ ;
  else
    if(  $\forall S \in R \exists i \in Range(region\_reaching\_defs\_copy(region\_cnt))$  such that  $S \subseteq i$ ) then
       $region\_table(R) := N$ ;
end;
let  $region(n) := region\_table$ ;

```

Fig. 5. Algorithm 2: Compute $region(n)$

3 Implementation

We implemented our Type Invariance Region Analysis (TIRA) framework in Vortex [11], an optimizing OOP back-end with front-ends for Cecil, Modula-3, C++, and Java. We implemented our framework as the last optimizing pass. The optimizations that precede our analysis are: Static Class Analysis [9], Iterative Class Analysis [10] [20], Class Hierarchy Analysis [12], Inlining [9], Splitting [10], and Common Subexpression Elimination [5]. The rationale for placing our analysis last is that we will optimize only those message sends that the earlier optimizations could not remove. Moreover, placing it after inlining gives our framework potentially larger regions to work with. Because our optimization uses reaching definitions of values rather than types, we do not need an interprocedural class analysis. We also focus on optimizing regions with more than one call-site because a region with one call-site is equivalent to an inline cache or run-time type test⁷.

In figure 2 $region(Alist)$ is defined after the statement $Alist := arg1$. We call this statement the *region entry point*. During code generation, we generate

⁷ A region with one call-site inside a loop may benefit, but most loops in our benchmarks were not implemented inline. Instead, they were implemented with a closure passed into a loop method.

run-time code to perform the *region type test* at the region entry point. We also generate static message sends for each region call-site.

Each region has a *region cache* variable that stores the most recent type binding for n in $region(n)$. Each region call-site has a cache that stores the most recent message target. At the region entry point, we generate a comparison between the type of n and the type stored in the region cache. Whenever the type binding for n differs from the region cache, we perform a dynamic dispatch for each region call-site. When n does not differ, it is safe to reuse the cached message targets for each region call-site. Figure 6 illustrates the run-time code for singly-dispatched regions. If the region consists of multiple-dispatched methods, a region cache for each dispatching argument is required.

```

OOP X;                                     /*assume some object X*/
static OOP_MAP * REGION_CACHE_TYPE = NULL; /*region cache*/
static void (*CALL_SITE_CACHE_1) (ARG_LIST); /*call-site cache 1*/
...
static void (*CALL_SITE_CACHE_N) (ARG_LIST); /*call-site cache N*/
/**** The code below is generated at the region entry point ****/
OOP_Map * Xtype = X->type();
if ( Xtype != REGION_CACHE_TYPE ){
    CALL_SITE_CACHE_1 = methodLookup(selector_1); /*lookup method 1*/
    ...
    CALL_SITE_CACHE_N = methodLookup(selector_N); /*lookup method N*/
    REGION_CACHE_TYPE = Xtype;          /*set region cache to new type*/
}
...
/***** Region call-sites below *****/
STATIC_SEND((*CALL_SITE_CACHE_1)(args)); /*directly call cached method 1*/
...
STATIC_SEND((*CALL_SITE_CACHE_N)(args)); /*directly call cached method N*/

```

Fig. 6. TIRA Run-Time Pseudo-code

3.1 Performance Evaluation

Table 1 summarizes our results⁸. The *Regions Detected* column breaks down the number of regions and sizes (e.g., number of call-sites) found by our framework. The *Regions Executed* column lists the number of regions and their sizes that were used at run-time. The *Hit Rate* column measures region cache performance. When the region cache misses, our run-time must perform method lookups for each region call-site. On a cache hit, each call-site directly invokes the method

⁸ Benchmarking was performed on an unloaded 333 MHz UltraSPARC 5/10 workstation with 256MB of RAM.

stored in its call-site cache. The *Type Tests Avoided* column lists the number of receiver type comparisons avoided based on the region size. *Time Base* is the execution time for the benchmark compiled with the default full optimization level in Vortex. *Time TIRA* is the execution time for the benchmark compiled with full optimization and our TIRA framework.

The majority of the regions detected were found in Cecil's standard library⁹. This probably explains the high cache hit rates as library methods are generally invoked at the end of a polymorphic call chain. This reveals a nice application for TIRA. TIRA can be used to optimize separately compiled libraries when no other type information is available.

4 Conclusion

In this work we developed a framework that augments existing intra and inter-procedural type analyses for OOPLs. Because our analysis works off of reaching definition values rather than types, our analysis can optimize message sends in which no available type information is present. This is especially useful for the separate compilation of library routines where the types of the incoming formal parameters are statically unknown but fixed at run-time. Next, we demonstrated in our benchmarks that most of the messages that we optimized had a single receiver type. This is evident with the high region cache hit rates and large regions found. Finally, we used our type invariant region analysis to remove redundant receiver type tests and improve performance. Average performance improvement was 9% with 18% in the best case.

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⁹ All 52 of deltablue's regions were in the library and 275 were in the library for the other four benchmarks.

Benchmark Performance									
Name	Description	Lines ^a	Regions Detected (size:qty)	Regions Executed (size:qty)	Hit Rate (%)	Type Tests Avoided (size:qty)	Time Base	Time TIRA	Speed up (%)
delta-blue	Constraint solver		2:36 3:12 4:3 5:1	2:2 3:1 4:0 5:1		2:14,624 3:264 5:5,764			
<i>total</i>		650	52	4	99	20,652	147ms	140ms	5.00
instr-sched	Global MIPS instruction scheduler		2:197 3:42 4:24 5:6 6:4 7:2 8:1	2:5 3:2 4:1 5:1 6:0 7:1 8:0		2:17,465 3:10,578 4:510 5:56,532 7:86,616			
<i>total</i>		2.4K	276	10	98	171,701	2.24s	2.11s	6.16
type-check	Typechecker for old Cecil type system		2:208 3:46 4:23 5:6 6:4 7:2 8:1	2:40 3:14 4:3 5:3 6:1 7:2 8:0		2:296,588 3:42,822 4:5,304 5:228,996 6:10 7:99,096			
<i>total</i>		20K	290	63	97	672,649	16s	14s	14.29
new-tc	Typechecker for new Cecil type system		2:216 3:48 4:25 5:6 6:5 7:2 8:1	2:43 3:14 4:6 5:3 6:2 7:2 8:0		2:342,489 3:79,352 4:43,551 5:72,904 6:9,965 7:190,434			
<i>total</i>		23.5K	303	70	97	738,695	13s	11s	18.18
comp	Old version of the Vortex Compiler		2:291 3:63 4:29 5:6 6:5 7:3 8:1 10:1	2:52 3:16 4:6 5:3 6:1 7:2 8:0 10:0		2:6,043,952 3:323,680 4:85,089 5:7,758,632 6:39,415 7:3,210,312			
<i>total</i>		50K	399	80	90	17,461,080	566s	552s	2.57

Table 1. Performance Evaluation

^a Does not include an 11,000 line standard library.

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