Abstract

The support for generic programming in modern object-oriented programming languages is awkward and lacks desirable expressive power. We introduce an expressive genericity mechanism that adds expressive power and strengthens static checking, while remaining lightweight and simple in common use cases. Like type classes and concepts, the mechanism allows existing types to model type constraints retroactively. For expressive power, we expose models as named constructs that can be defined and selected explicitly to witness constraints; in common uses of genericity, however, types implicitly witness constraints without additional programmer effort. Models are integrated into the object-oriented style, with features like model generics, model-dependent types, model enrichment, model multimethods, constraint entailment, model inheritance, and existential quantification further extending expressive power in an object-oriented setting. We introduce the new genericity features and show that common generic programming idioms, including current generic libraries, can be expressed more precisely and concisely. The static semantics of the mechanism and a proof of a key decidability property are provided.

1 Introduction

Generic programming provides the means to express algorithms and data structures in an abstract, adaptable, and interoperable form. Specifically, genericity mechanisms allow polymorphic code to apply to different types, improving modularity and reuse. Despite decades of work on genericity mechanisms, current OO languages still offer an unsatisfactory tradeoff between expressiveness and usability. These languages do not provide a design that coherently integrates desirable features—particularly, retroactive extension and dynamic dispatch. In practice, existing genericity mechanisms force developers to circumvent limitations in expressivity by using awkward, heavyweight design patterns and idioms.

The key question is how to expose the operations of type parameters in a type-safe, intuitive, and flexible manner within the OO paradigm. The following somewhat daunting Java signature for method Collections::sort illustrates the problem:

\[
\langle T \text{ extends } \text{Comparable}\rangle \langle? \text{ super } T\rangle \rangle \text{ void sort(List}\langle T\rangle \text{ l)}
\]

The subtyping constraint constrains a type parameter T using the Comparable interface, ensuring that type T is comparable to itself or to one of its supertypes. However, sort can only be used on a type T if that type argument is explicitly declared to implement the Comparable interface. This restriction of nominal subtyping is alleviated by structural constraints as introduced by CLU [Liskov et al. 1977, 1984] and applied elsewhere (e.g., [Day et al. 1995; Chambers 1992]), but a more fundamental limitation remains: items of type T cannot
class AbstractVertex
    <EdgeType extends AbstractEdge<EdgeType, ActualVertexType>,
    ActualVertexType extends AbstractVertex<EdgeType, ActualVertexType>> {...}

class AbstractEdge
    <ActualEdgeType extends AbstractEdge<ActualEdgeType, VertexType>,
    VertexType extends AbstractVertex<ActualEdgeType, VertexType>> {...}

Figure 1: Parameter clutter in generic code.

be sorted unless \( T \) has a `compareTo` operation to define the sort order. That limitation is addressed by `type classes` in Haskell [Wadler and Blott 1989]. Inspired by Haskell, efforts have been made to incorporate type classes into OO languages with language-level support [Siek and Lumsdaine 2011; Wehr and Thiemann 2011; Rust; Swift] and the Concept design pattern [Oliveira et al. 2010]. However, as we argue, these designs do not fully exploit what type classes and OO languages have to offer when united.

This paper introduces a new genericity mechanism, embodied in a new extension of Java called Genus. The genericity mechanism enhances expressive power, code reuse, and static type safety, while remaining lightweight and intuitive for the programmer in common use cases. Genus supports `models` as named constructs that can be defined and selected explicitly to witness constraints, even for primitive type arguments; however, in common uses of genericity, types implicitly witness constraints without additional programmer effort. The key novelty of models in Genus is their deep integration into the OO style, with features like model generics, model-dependent types, model enrichment, model multimethods, constraint entailment, model inheritance, and existential quantification further extending expressive power in an OO setting.

The paper compares Genus to other language designs; describes its implementation; shows that Genus enables safer, more concise code through experiments that use it to reimplement existing generic libraries; and presents performance measurements that show that a naive translation from Genus to Java yields acceptable performance and that with simple optimizations, Genus can offer very good performance. The paper also gives a formal static semantics for a core version of Genus, and show that termination of subtyping and default model resolution hold under reasonable syntactic restrictions.

2 The Need for Better Genericity

Prior work has explored various approaches to constrained genericity: subtyping constraints, structural matching, type classes, and design patterns. Each of these approaches has significant weaknesses.

The trouble with subtyping. Subtyping constraints are used in Java [Bracha et al. 1998], C# [Kennedy and Syme 2001; Emir et al. 2006], and other OO languages. In the presence of nominal subtyping, subtyping constraints are too inflexible: they can only be satisfied by classes explicitly declared to implement the constraint. Structural subtyping and matching mechanisms (e.g., [Liskov et al. 1984; Chambers 1992; Day et al. 1995; Myers et al. 1997]) do not require an explicit declaration that a constraint is satisfied, but still require that the relevant operations exist, with conformant signatures. Instead, we want retroactive modeling, in which a `model` (such as a type class instance [Wadler and Blott 1989]) can define how an existing type satisfies a constraint that it was not planned to satisfy ahead of time.

Subtyping constraints, especially when F-bounded [Canning et al. 1989], also tend to lead to complex code when multiple type parameters are needed. For example, Figure 1 shows a simplification of the signatures of the classes `AbstractVertex` and `AbstractEdge` in the FindBugs project [findbugs-release]. The
class TreeSet<T> implements Set<T> {
    TreeSet(Comparator<? super T> comparator) {...} ...
}

interface Comparator<T> { int compare(T o1, T o2); }

Figure 2: Concept design pattern.

vertex and the edge types of a graph have a mutual dependency that is reflected in the signatures in an unpleasantly complex way (See Figure 3 for our approach).

Concept design pattern. Presumably because of these limitations, the standard Java libraries mostly do not use constraints on the parameters of generic classes in the manner originally envisioned [Bracha et al. 1998]. Instead, they use a version of the Concept design pattern [Myers 1995] in which operations needed by parameter types are provided as arguments to constructors. For instance, a constructor of TreeSet, a class in the Java collections framework, accepts an object of the Comparator class (Figure 2). The compare operation is provided by this object rather than by T itself.

This design pattern provides missing flexibility, but adds new problems. First, a comparator object must be created even when the underlying type has a comparison operation. Second, because the model for Comparator is an ordinary (first-class) object, it is hard to specialize or optimize particular instantiations of generic code. Third, there is no static checking that two TreeSets use the same ordering; if an algorithm relies on the element ordering in two TreeSets being the same, the programmer may be in for a shock.

In another variant of the design pattern, used in the C++ STL [Musser et al. 2001], an extra parameter for the class of the comparator distinguishes instantiations that use different models. However, this approach is more awkward than the Comparator object approach. Even the common case, in which the parameter type has exactly the needed operations, is just as heavyweight as when an arbitrary, different operation is substituted.

Type classes and concepts. The limitations of subtyping constraints have led to recent research on adapting type classes to OO languages to achieve retroactive modeling [Siek and Lumsdaine 2011]. However, type classes have limitations: first, constraint satisfaction must be uniquely witnessed, and second, their models define how to adapt a single type, whereas in a language with subtyping, each adapted type in general represents all of its subtypes.

No existing approach addresses the first limitation, but an attempt is made by JavaGI [Wehr and Thiemann 2011] to fit subtyping polymorphism and dynamic dispatch into constrained genericity. As we will argue ($\S$5.1), JavaGI’s limited dynamic dispatch makes certain constraints hard to express, and interactions between subtyping and constraint handling make type checking subject to nontermination.

Beyond dynamic dispatch, it is important for OO programming that extensibility applies to models as well. The essence of OO programming is that new behavior can be added later in a modular way; we consider this post-factum enrichment of models to be a requirement.

Goals. What is wanted is a genericity mechanism with multiple features: retroactive modeling, a lightweight implicit approach for the common case, multiparameter type constraints, non-unique constraint satisfaction with dynamic, extensible models, and model-dependent types. The mechanism should support modular compilation. It should be possible to implement the mechanism efficiently; in particular, an efficient implementation should limit the use of wrapper objects and should be able to specialize generic code to particular type arguments—especially, to primitive types. Genus meets all of these goals. We have tried not only to address the immediate problems with generics seen in current OO languages, but also to take further steps, adding features that support the style of programming that we expect will evolve when generics are easier to use than they are now.
// A multiparameter constraint
constraint GraphLike[V,E] {
    Iterable[E] V.outgoingEdges();
    Iterable[E] V.incomingEdges();
    V E.source();
    V E.sink();
}

constraint OrdRing[T] extends Comparable[T] {
    // static methods
    static T T.zero();
    static T T.one();
    T T.plus(T that);
    T T.times(T that);
}

Figure 3: Constraints GraphLike and OrdRing.

3 Type Constraints in Genus

3.1 Type Constraints as Predicates

Instead of constraining types with subtyping, Genus uses explicit type constraints similar to type classes. For example, the constraint:

constraint Eq[T] {
    boolean equals(T other);
}

requires that type T have an equals method. Although this constraint looks like a Java interface, it is really a predicate on types, like a (multiparameter) type class in Haskell [Peyton-Jones et al. 1997]. We do not call constraints “type classes” because there are differences and because the name “class” is already taken in the OO setting.

Generic code can require that actual type parameters satisfy constraints. For example, here is the Set interface in Genus (simplified):

interface Set[T where Eq[T]] { ... }

The where clause “where Eq[T]” establishes the ability to test equality on type T within the scope of Set. Consequently, an instantiation of Set needs a witness that Eq is satisfied by the type argument. In Genus, such witnesses come in the form of models. Models are either implicitly chosen by the compiler or explicitly supplied by the programmer.

Multiparameter constraints. A constraint may be a predicate over multiple types. Figure 3 contains an example in which a constraint GraphLike[V,E] declares graph operations that should be satisfied by any pair of types [V,E] representing vertices and edges of a graph. In a multiparameter constraint, methods must explicitly declare receiver types (V or E in this case). Every operation in this constraint mentions both V and E; none of the operations really belongs to any single type. The ability to group related types and operations into a single constraint leads to code that is more modular and more readable than that in Figure 1.

Prerequisite constraints. A constraint can have other constraints as its prerequisites. For example, Eq[T] is a prerequisite constraint of Comparable[T]:

constraint Comparable[T] extends Eq[T] {
    int compareTo(T other);
}

To satisfy a constraint, its prerequisite constraints must also be satisfied. Therefore, the satisfaction of a constraint entails the satisfaction of its prerequisites. For example, the Genus version of the TreeSet class from Figure 2 looks as follows:

1We denote Genus type parameters using square brackets, to distinguish Genus examples from those written in other languages (especially, Java).
Map[V,W] SSSP[V,E,W](V s)
    where GraphLike[V,E], Weighted[E,W],
        OrdRing[W], Hashable[V] {
        TreeMap[W,V] frontier = new TreeMap[W,V]();
        Map[V,W] distances = new HashMap[V,W]();
        distances.put(s, W.one()); frontier.put(W.one(), s);
        while (frontier.size() > 0) {
            V v = frontier.pollFirstEntry().getValue();
            for (E vu : v.outgoingEdges()) {
                V u = vu.sink();
                W weight = distances.get(v).times(vu.weight());
                if (!distances.containsKey(u) ||
                    weight.compareTo(distance.get(u)) < 0) {
                    frontier.put(weight, u);
                    distances.put(u, weight);
                }
            }
        }
        return distances; }

Figure 4: A highly generic method for Dijkstra’s single-source shortest-path algorithm. Definitions of Weighted and Hashable are omitted. Ordering and composition of distances are generalized to an ordered ring. (A more robust implementation might consider using a priority queue instead of TreeMap.)

class TreeSet[T where Comparable[T]] implements Set[T] {
    TreeSet() { ... } ...
}

The type Set[T] in the definition of TreeSet is well-formed because its constraint Eq[T] is entailed by the constraint Comparable[T].

Static constraint members. Constraints can require that a type provide static methods, indicated by using the keyword static in the method declaration. In Figure 3, constraint OrdRing specifies a static method (zero) that returns the identity of the operation plus.

All types T are also automatically equipped with a static method T.default() that produces the default value for type T. This method is called, for instance, to initialize the elements of an array of type T[], as in the following example:

class ArrayList[T] implements List[T] {
    T[] arr;
    ArrayList() {
        arr = new T[INITIAL_SIZE]; // Calls T.default()
    }
    ...
}

The ability to create an array of type T[] is often missed in Java.

3.2 Prescribing Constraints Using Where Clauses

Where-clause constraints enable generic algorithms, such as the version of Dijkstra’s shortest-path algorithm in Figure 4, generalized to ordered rings. The where clause of SSSP requires only that the type arguments satisfy their respective constraints—no subtype relationship is needed.

Where-clause constraints endow typing contexts with assumptions that the constraints are satisfied. So the code of SSSP can make method calls like vu.sink() and W.one(). Note that the where clause may be placed after the formal parameters as in CLU; this notation is just syntactic sugar for placing it between the brackets.

\(^2\)The usual behavior is achieved if plus is min, times is +, and one is 0.
Unlike Java extends clauses, a where clause is not attached to a particular parameter. It can include multiple constraints, separated by commas. Each constraint requires a corresponding model to be provided when the generic is instantiated. To allow models to be identified unambiguously in generic code, each such constraint in the where clause may be explicitly named as a model variable.

Another difference from Java extends clauses is that a where clause may be used without introducing a type parameter. For example, consider the remove method of List. Expressive power is gained if its caller can specify the notion of equality to be used, rather than requiring List itself to have an intrinsic notion of equality. Genus supports this genericity by allowing a constraint `Eq[E]` to be attached to remove:

```java
interface List[E] {
    boolean remove(E e) where Eq[E]; ...
}
```

We call this feature model genericity.

### 3.3 Witnessing Constraints Using Models

As mentioned, generic instantiations require witnesses that their constraints are satisfied. In Genus, witnesses are provided by models. Models can be inferred—a process we call default model resolution—or specified explicitly, offering both convenience in common cases and expressivity when needed. We start with the use of models and leave the definition of models until §4.

**Using default models.** It is often clear from the context which models should be used to instantiate a generic. For instance, the `Set[T]` interface in the TreeSet example (§3.1) requires no further annotation to specify a model for `Eq[T]`, because the model can be uniquely resolved to the one promised by `Comparable[T]`.

Another common case is that the underlying type already has the required operations. This case is especially likely when classes are designed to support popular operations; having to supply models explicitly in this case would be a nuisance. Therefore, Genus allows types to structurally conform to constraints. When the methods of a type have the same names as the operations required by a constraint, and also have conformant signatures, the type automatically generates a natural model that witnesses the constraint. For example,³ the type `Set[String]` means a Set that distinguishes strings using `String`'s built-in `equals` method. Thus, the common case in which types provide exactly the operations required by constraints is simple and intuitive. In turn, programmers have an incentive to standardize the names and signatures of popular operations.

Genus supports using primitive types as type arguments, and provides natural models for them that contain common methods. For example, a natural model for `Comparable[int]` exists, so types like `TreeSet[int]` that need that model can be used directly.

Default models can be used to instantiate any generic—not just generic classes. For example, consider this sort method:

```java
void sort[T](List[T] l) where Comparable[T] { ... }
```

The call `sort(x)`, where `x` is a `List[int]`, infers `int` both as the type argument and as the default model. Default model resolution, and more generally, type and model inference, are discussed further in §4.4 and §4.7.

**Using named models.** It is also possible to explicitly supply models to witness constraints. To do so, programmers use the with keyword followed by models for each of the where-clause constraints in the generic. These models can come from programmer-defined models (§4) or from model variables declared in where

---

³We assume throughout that the type `String` has methods “`boolean equals(String)`” and “`int compareTo(String)`.”
clauses (§3.2). For example, suppose model CIEq tests String equality in a case-insensitive manner. The type Set[String with CIEq] then describes a Set in which all strings are distinct without case-sensitivity. In fact, the type Set[String] is syntactic sugar for Set[String with String], in which the with clause is used to explicitly specify the natural model that String automatically generates for Eq[String].

A differentiating feature of our mechanism is that different models for Eq[String] can coexist in the same scope, allowing a generic class like Set, or a generic method, to be instantiated in more than one way in a scope:

```java
Set[String] s0 = ...;
Set[String with CIEq] s1 = ...;
s1 = s0; // illegal assignment: different types.
```

The ordering that an instantiation of Set uses for its elements is part of the type, rather than a purely dynamic argument passed to a constructor as in the Concept pattern. Therefore, the final assignment statement is a static type error. The type checker catches the error because the different models used in the two Set instantiations allow Sets using different notions of equality to be distinguished. The use of models in types is discussed further in §4.5.

It is also possible to express types using wildcard models; the type Set[String with ?] is a supertype of both Set[String] and Set[String with CIEq]. Wildcard models are actually syntactic sugar for existential quantification (§6).

### 4 Models

Models can be defined explicitly to allow a type to satisfy a constraint when the natural model is nonexistent or undesirable. For example, the case-insensitive string equality model CIEq can be defined concisely:

```java
model CIEq for Eq[String] {
    bool equals(String str) {
        return equalsIgnoreCase(str);
    }
}
```

Furthermore, a model for case-insensitive String ordering might be defined by reusing CIEq via model inheritance, to witness the prerequisite constraint Eq[String]:

```java
model CICmp for Comparable[String] extends CIEq {
    int compareTo(String str) {
        return compareToIgnoreCase(str);
    }
}
```

It is also possible for CICmp to satisfy Eq by defining its own equals method. Model inheritance is revisited in §5.3.

Models are immutable: they provide method implementations but do not have any instance variables. Models need not have global scope; modularity is achieved through the Java namespace mechanism. Similarly, models can be nested inside classes and are subject to the usual visibility rules.

#### 4.1 Models as Expanders

Operations provided by models can be invoked directly, providing the functionality of expanders [Warth et al. 2006]. For example, the call "x". (CIEq.equals)("X") uses CIEq as the expander to test equality of two strings while ignoring case. Natural models can similarly be selected explicitly using the type name: "x". (String.equals)("X").
Using models as expanders is an integral part of our genericity mechanism: the operations promised by where-clause constraints are invoked using expanders. In Figure 4, if we named the where-clause constraint GraphLike[V,E] with model variable g, the call vu.sink() would be sugar for vu.(g.sink)() with g being the expander. In this case, the expander can be elided because it can be inferred via default model resolution (§4.4).

4.2 Parameterized Models

Model definitions can be generic: they can be parameterized with type parameters and where-clause constraints. For example, model ArrayListDeepCopy (Figure 5) gives a naive implementation of deep-copying ArrayLists. It is generic with respect to the element type E, but requires E to be cloneable.

As another example, we can exploit model parameterization to implement the transpose of any graph. In Figure 6, the DualGraph model is itself a model for GraphLike[V,E], and is parameterized by another model for GraphLike[V,E] (named g). It represents the transpose of graph g by reversing its edge orientations.

4.3 Non-Uniquely Witnessing Constraints

Previous languages with flexible type constraints, such as Haskell, JavaGI, and G, require that witnesses be unique at generic instantiations, whether witnesses are scoped globally or lexically. By contrast, Genus allows multiple models witnessing a given constraint instantiation to coexist in the same context. This flexibility increases expressive power.

For example, consider Kosaraju’s algorithm for finding strongly connected components in a directed graph [Aho et al. 1983]. It performs two depth-first searches, one following edges forward, and the other on the transposed graph, following edges backward. We would like to reuse the same generic depth-first-search algorithm on the same graph data structure for both traversals.
In Figure 6, the where clause of SCC introduces into the context a model for GraphLike[V,E], denoted by model variable g. Using the DualGraph model, the algorithm code can then perform both forward and backward traversals. It instantiates DFIterator, an iterator class for depth-first traversal, twice, with the original graph model g and with the transposed one. Being able to use two different models to witness the same constraint instantiation in SCC enables more code reuse. The highlighted with clauses can be safely elided, which brings us to default model resolution.

### 4.4 Resolving Default Models

In Genus, the omission of a with clause triggers default model resolution. Default model resolution is based on the following four ways in which models are enabled as potential default choices. First, types automatically generate natural models when they structurally conform to constraints. Natural models, when they exist, are always enabled as default candidates. Second, a where-clause constraint enables a model within the scope of the generic to which the where clause is attached. For example, in method SCC in Figure 6 the where clause enables a model as a default candidate for GraphLike[V,E] within SCC. Third, a use declaration, e.g.,

```java
use ArrayListDeepCopy;
```

enables the specified model as a potential default way to clone ArrayLists in the compilation unit in which the declaration resides. Fourth, a model itself is enabled as a potential default model within its definition.

Default model resolution works as follows:

1. If just one model for the constraint is enabled, it becomes the default model.
2. If more than one model is enabled, programmer intent is ambiguous. In this case, Genus requires that programmers make their intent explicit using a with clause. Omitting the with clause is a static error in this case.
3. If no model is explicitly enabled, but there is in scope a single model for the constraint, that model becomes the default model for the constraint.

Resolution for an elided expander in a method call works similarly. The only difference is that instead of searching for a model that witnesses a constraint, the compiler searches for a model that contains a method applicable to the given call. In typical use, this would be the natural model.

These rules for default models make generics and expanders easy to use in the common cases; in the less common cases where there is some ambiguity about which model to use, they force the programmer to be explicit and thereby help prevent hard-to-debug selection of the wrong model.

Letting each compilation unit choose its own default models is more flexible and concise than using Scala implicits, where a type-class instance can only be designated as implicit at the place where it is defined, and implicit definitions are then imported into the scope, with a complex process used to find the most specific implicit among those imported [Odersky 2014]. We aim for simpler rules.

Genus also achieves the conciseness of Haskell type classes because uniquely satisfying models are allowed to witness constraints without being enabled, just as a unique type class instance in Haskell satisfies its type class without further declarations. But natural models make the mechanism lighter-weight than in Haskell, and the ability to have multiple models adds expressive power (as in the SCC example in Figure 6).

### 4.5 Models in Types

Section 3.3 introduced the ability to instantiate generic types with models, which become part of the type (i.e., model-dependent types). Type safety benefits from being able to distinguish instantiations that use different models.
class TreeSet[T] implements Set[T with Comparable[T] c] {
    TreeSet() {...}
    void addAll(Collection?[extends T] src) {
        if (src instanceof TreeSet?[extends T with Comparable[T] c]) {
            addFromSorted((TreeSet?[extends T with Comparable[T] c]) src);
        } else {...}
    }
    void addFromSorted(TreeSet?[extends T with Comparable[T] c] src) {
        ...
        // specialized code in virtue of the same ordering in src and this
    }
}

Figure 7: TreeSet in Genus. Highlighted code is inferred if omitted.

The addFromSorted method in TreeSet (Figure 7) adds all elements in the source TreeSet to this one. Its signature requires that the source TreeSet and this one use the same ordering. So a TreeSet with a different ordering cannot be accidentally passed to this method, avoiding a run-time exception.

Including the choice of model as part of the type is unusual, perhaps because it could increase annotation burden. Models are not part of types in the Concept design pattern (e.g., as realized in Scala [Oliveira et al. 2010]), because type class instances are not part of instantiated types. Ψ [Siek and Lumsdaine 2011] allows multiple models for the same constraint to be defined in one program (albeit only one in any lexical scope), yet neither at compile time nor at run time does it distinguish generic instantiations with distinct models. This raises potential safety issues when different modules interoperate.

In Genus, the concern about annotation burden is addressed by default models. For example, the type TreeSet?[extends T] in Figure 7 is implicitly instantiated with the model introduced by the where clause (via constraint entailment, §5.2). By contrast, Scala implicits work for method parameters, but not for type parameters of generic classes.

4.6 Models at Run Time

Unlike Java, whose type system is designed to support implementing generics via erasure, Genus makes models and type arguments available at run time. Genus allows testing the type of an object from a parameterized class at run time, like the instanceof test and the type cast in Figure 7.

Refiability creates opportunities for optimization. For example, consider TreeSet’s implementation of the addAll method required by the Collection interface. In general, an implementation cannot rely on seeing the elements in the order expected by the destination collection, so for each element in the source collection, it must traverse the destination TreeSet to find the correct position. However, if both collections use the same ordering, the merge can be done in a more asymptotically efficient way by calling the specialized method addFromSorted.

4.7 Default Model Resolution: Algorithmic Issues

Recursive resolution of default models. Default model resolution is especially powerful because it supports recursive reasoning. For example, the use declaration in §4.4 is syntactic sugar for the following parameterized declaration:


The default model candidacy of ArrayListDeepCopy is valid for cloning objects of any instantiated ArrayList type, provided that the element type satisfies Cloneable too. Indeed, when the compiler investigates the
use of `ArrayListDeepCopy` to clone `ArrayList[Foo]`, it creates a subgoal to resolve the default model for `Cloneable[Foo]`. If this subgoal fails to be resolved, `ArrayListDeepCopy` is not considered as a candidate.

Recursive resolution may not terminate without additional restrictions. As an example, the declaration “use `DualGraph`;” is illegal because its recursive quest for a model of the same constraint causes resolution to cycle. The issue is addressed in §9 and the technical report [Zhang et al. 2015] by imposing syntactic restrictions.

When a use declaration is rejected by the compiler for violating the restrictions, the programmer always has the workaround of explicitly selecting the model. By contrast, the inability to do so in Haskell or JavaGI makes it impossible to have a model like `DualGraph` in these languages.

**Unification vs. default model resolution.** Since Genus uses models in types, it is possible for models to be inferred via unification when they are elided. This inference potentially raises confusion with default model resolution.

Genus distinguishes between two kinds of where-clause constraints. Constraints for which the model is required by a parameterized type, such as `Eq[T]` in the declaration `void f[T where Eq[T]](Set[T] x)`, are called *intrinsic constraints*, because the `Set` must itself hold the corresponding model. By contrast, a constraint like `Printable[T]` in the declaration `void g[T where Printable[T]](List[T] x)` is *extrinsic* because `List[T]` has no such constraint on `T`.

Inference in Genus works by first solving for type parameters and intrinsic constraints via unification, and only then resolving default models for extrinsic constraints. To keep the semantics simple, Genus does not use default model availability to guide unification, and it requires extrinsic where-clause constraints to be written to the right of intrinsic ones. Nevertheless, it is always possible for programmers to explicitly specify intent.

### 4.8 Constraints/Models vs. Interfaces/Objects

The relationship between models and constraints is similar to that between objects and interfaces. Indeed, the Concept pattern can be viewed as using objects to implement models, and JavaGI extends interfaces to encode constraints. In contrast, Genus draws a distinction between the two, treating models as second-class values that cannot be stored in ordinary variables. This design choice has the following basis:

- Constraints are used in practice very differently from “ordinary” types, as evidenced by the nearly complete separation between *shapes* and *materials* seen in an analysis of a very large software base [Greenman et al. 2014]. In their parlance, interfaces or classes that encode multiparameter constraints (e.g., `GraphLike`) or constraints requiring binary operations (e.g., `Comparable`) are shapes, while ordinary types (e.g., `Set`) are materials. Muddling the two may give rise to nontermination (§9).
- Because models are not full-fledged objects, generic code can easily be specialized to particular using contexts.
- Because model expressions can be used in types, Genus has dependent types; however, making models second-class and immutable simplifies the type system and avoids undecidability.

### 5 Making Models Object-Oriented

#### 5.1 Dynamic Dispatching and Enrichment

In OO programs, subclasses are introduced to specialize the behavior offered by their superclasses. In Genus, models define part of the behavior of objects, so models too should support specialization. Therefore, a model in Genus may include not only method definitions for the base type, but also methods defining more specific behavior for subtypes. These methods can be dispatched *dynamically* by code both inside
and outside model declarations. Dynamic dispatch takes place not only on the receiver, but also on method arguments of the manipulated types. The expressive power of dynamic dispatch is key to OO programming [Aldrich 2013], and multiple dispatch is particularly important for binary operations, which are typically encoded as constraints. Our approach differs in this way from $G$ and Scala, which do not support dynamic dispatch on model operations.

For example, model ShapeIntersect in Figure 8 gives multiple definitions of intersect, varying in their expected argument types. In a context where the model is selected, a call to intersect on two objects statically typed as Shape will resolve at run time to the most specific method definition in the model. In JavaGI, multiple dispatch on intersect is impossible, because its dispatch is based on “self” types [Bruce et al. 1998], while the argument types (including receiver) as well as the return type of an intersect implementation do not necessarily have to be the same.

Existing OO type hierarchies are often extended with new subclasses in ways not predicted by their designers. Genus provides model enrichment to allow models to be extended in a modular way, in sync with how class hierarchies are extended; here we apply the idea of open classes [Clifton et al. 2000] to models. For example, if Triangle is later introduced to the Shape hierarchy, the model can then be separately enriched, as shown in the enrich declaration in Figure 8.

Model multimethods and model enrichment create the same challenge for modular type checking that is seen with other extensible OO mechanisms. For instance, if two modules separately enrich ShapeIntersect, these enrichments may conflict. Like Relaxed MultiJava [Millstein et al. 2003], Genus can prevent such errors with a load-time check that there is a unique best method definition for every method invocation, obtaining mostly modular type checking and fully modular compilation. The check can be performed soundly, assuming load-time access to the entire program. If a program loads new code dynamically, the check must be performed at the time of dynamic loading.

5.2 Constraint Entailment

As seen earlier (§3.1), a constraint entails its prerequisite constraints. In general, a model may be used as a witness not just for the constraint it is declared for, but also for any constraints entailed by the declared constraint. For example, a model for Comparable[Shape] can be used to witness Eq[Shape].

A second way that one constraint can entail another is through variance on constraint parameters. For example, since in constraint Eq the type parameter only occurs in contravariant positions, a model for Eq[Shape] may also be soundly used as a model for Eq[Circle]. It is also possible, though less common, to use a model to witness constraints for supertypes, via covariance. Variance is inferred automatically by the compiler, with bivariance downgraded to contravariance.

A model enabled for some constraint in one of the four ways discussed in §4.4 is also enabled for its prerequisite constraints and constraints that can be entailed via contravariance. Accommodating subtyping extends the expressivity of default model resolution, but poses new challenges for termination. In §9, we
show that encoding “shape” types (in the sense of Greenman et al. [2014]) as constraints helps ensure termination.

5.3 Model Inheritance

Code reuse among models can be achieved through model inheritance, signified by an extends clause (e.g., model CICmp in §4). Unlike an extends clause in a class or constraint definition, which creates an is-a relationship between a subclass and its superclass or a constraint and its prerequisite constraint, an extends clause in a model definition is merely for code reuse. The inheriting model inherits all method definitions with compatible signatures available in the inherited model. The inheriting model can also override these inherited definitions.

Model inheritance provides a means to derive models that are otherwise rejected by constraint entailment. For example, the model ShapeIntersect (Figure 8) soundly witnesses the same constraint for Rectangle, because the selected method definitions have compatible signatures, even though Intersectable is invariant with respect to its type parameter. The specialization to Rectangle can be performed succinctly using model inheritance, with the benefit of a more precise result type when two rectangles are intersected:

```java
model RectangleIntersect for Intersectable[Rectangle]
  extends ShapeIntersect { }
```

6 Use-Site Genericity

Java’s wildcard mechanism [Torgersen et al. 2004] is in essence a limited form of existential quantification. Existentials enable genericity at use sites. For example, a Java method with return type List<? extends Printable> can be used by generic calling code that is able to print list elements even when the type of the elements is unknown to the calling code. The use-site genericity mechanism of Genus generalizes this idea while escaping some limitations of Java wildcards. Below we sketch the mechanism.

6.1 Existential Types

Using subtype-bounded existential quantification, the Java type List<? extends Printable> might be written more type-theoretically as \( \exists U \leq \text{Printable.List}[U] \). Genus extends this idea to constraints. An existential type in Genus is signified by prefixing a quantified type with type parameters and/or where-clause constraints. For example, if Printable is a constraint, the Genus type corresponding to the Java type above is \([\text{some } U \text{ where } \text{Printable}[U]]\text{List}[U] \). The initial brackets introduce a use-site type parameter \( U \) and a model for the given constraint, which are in scope in the quantified type; the syntax emphasizes the connection between existential and universal quantification.

The presence of prefixed parameters in existential types gives the programmer control over the existential binding point, in contrast to Java wildcard types where binding is always at the generic type in which the wildcard is used as a type argument. For example, no Java type can express \( \exists U.\text{List}[\text{List}[U]] \), meaning a homogeneous collection of lists in which each list is parameterized by the same unknown type. This type is easily expressed in Genus as \([\text{some } U]\text{List}[\text{List}[U]] \).

Genus also offers convenient syntactic sugar for common uses of existential types. A single-parameter constraint can be used as sugar for an existential type: e.g., Printable, used as a type, is sugar for \([\text{some } U \text{ where } \text{Printable}[U]]U \), allowing a value of any printable type. The wildcard syntax List[?] represents an existential type, with the binding point the same as in the Java equivalent. The type with a wildcard model Set[String with ?] is sugar for \([\text{some Eq[String]} m]\text{Set}[\text{String with } m] \).
some T where Comparable[T] List[T] f () {
    return new ArrayList[String]();
}

sort(f());

[U] (List[U] l) where Comparable[U] = f(); // bind U
l.first().compareTo(l.last()); // U is comparable
U[] a = new U[64]; // use run-time info about U
l = new List[U](); // new list, same U

Figure 9: Working with existential quantification.

Subtyping and coercion. Genus draws a distinction between subtyping and coercion involving existential types. Coercion may induce extra computation (i.e., existential packing) and can be context-dependent (i.e., default model resolution), while subtyping cannot. For example, the return expression in Figure 9 type-checks not because ArrayList[String] is a subtype of the existential return type, but because of coercion, which works by packing together a value of type ArrayList[String] with a model for Comparable[String] (in this case, the natural model) into a single value. The semantics of subtyping involving where-clause-quantified existential types is designed in a way that makes it easy for programmers to reason about subtyping and joining types.

Capture conversion. In Java, wildcards in the type of an expression are instantiated as fresh identifiers when the expression is type-checked, a process called capture conversion [Gosling et al. 2005]. Genus extends this idea to constraints: in addition to fresh type variables, capture conversion generates fresh models for where-clause constraints, and enables them in the current scope.

For example, at line 1 in Figure 9, when the call to sort (defined in §3.3) is type-checked, the type of the call f() is capture-converted to List[#T], where #T is the fresh type variable that capture conversion generates for T, and a model for Comparable[#T] becomes enabled in the current context. Subsequently, the type argument to sort is inferred as #T, and the default model for Comparable[#T] resolves to the freshly generated model.

6.2 Explicit Local Binding

Capture conversion is convenient but not expressive enough. Consider a Java object typed as List<? extends Comparable>. The programmer might intend the elements of this homogeneous list to be comparable to one another, but comparisons to anything other than null do not type-check.

The awkwardness is addressed in Genus by explicit local binding of existentially quantified type variables and where-clause constraints, giving them names that can be used directly in the local context. An example of this mechanism is found at line 2 in Figure 9. The type variable U can be used as a full-fledged type in the remainder of the scope.

As its syntax suggests, explicit local binding can be viewed as introducing an inlined generic method encompassing subsequent code. Indeed, it operates under the same rules as universally quantified code. For example, the where clause at line 2 enables a new model so that values of type U can be compared at line 3. Also, locally bound type variables are likewise reifiable (line 4). Moreover, the binding at line 2 is type-checked using the usual inference algorithm to solve for U and for the model for Comparable[U]: per §4.7, the former is inferred via unification and the latter via default model resolution—it is an extrinsic constraint. Soundness is maintained by ensuring that 1 is initialized upon declaration and that assignments to the variable preserve the meaning of U.
7 Implementation

We have built a partial implementation of the Genus language in Java. The implementation consists of about 23,000 lines of code, extending version 2.6.1 of the Polyglot compiler framework [Nystrom et al. 2003]. We have an essentially complete implementation of all type checking and inference features. Code generation works by translating to Java 5 code, relying on a Java compiler as a back end. Code generation is less complete than type checking but also less interesting; however, the compiler can compile the benchmarks of §8, which use generics in nontrivial ways. The current compiler implementation type-checks but does not generate code for multimethod dispatch or for existentials; it does not yet specialize instantiations to particular type arguments.

7.1 Implementing Constraints and Models

Constraints and models in Genus code are translated to parameterized interfaces and classes in Java. For example, the constraint \textbf{Comparable}[T] is translated to a parameterized Java interface \textbf{Comparable}<T,A$T> providing a method \texttt{compareTo} with the appropriate signature: \texttt{int compareTo(T,T)}. The extra type parameter A$T is explained in §7.3. Models are translated to Java classes that implement these constraint interfaces.

7.2 Implementing Generics

Parameterized Genus classes are translated to correspondingly parameterized Java classes. However, type arguments and models must be represented at run time. So extra arguments carrying this information are required by class constructors, and constructor bodies are extended to store these arguments as fields. For example, class \textbf{ArrayList} has a translated constructor with the signature shown in Figure 10. Parameterized methods and models are translated in a similar way by adding extra arguments representing type and model information.

7.3 Supporting Primitive Type Arguments

A challenge for efficient generics, especially with a JVM-based implementation, is how to avoid uniformly wrapping all primitives inside objects when primitive types are used as type arguments. Some wrapping is unavoidable, but from the standpoint of efficiency, the key is that when code parameterized on a type T is instantiated on a primitive type (e.g., int), the array type T[] should be represented exactly as an array of the primitive type (e.g., int[]), rather than a type like Integer[] in which every array element incurs the overhead of individualized memory management.
Our current implementation uses a homogeneous translation to support this efficiency; the model object (e.g., $T_{\text{model}}$ in Figure 10) for a type parameter $T$ provides all operations about $T[]$. The model object has the interface type $\text{ObjectModel}<$T, A$T>$, which specifies, via $A$T, the operations for creating and accessing arrays of (unboxed) $T$. For example, the type of the model object used to create an ArrayList[double] implements $\text{ObjectModel}$-Double, double[]$. All interfaces representing single-parameter constraints implicitly extend $\text{ObjectModel}<T, A$T$>$, so an $\text{ObjectModel}$ argument is usually needed only on generics that do not otherwise constrain their type parameters.

A more efficient approach to supporting primitive type arguments is to generate specialized code for primitive instantiations, as is done in C#. The design of Genus makes it straightforward to implement particular instantiations with specialized code.

8 Evaluation

8.1 Porting Java Collections Framework to Genus

To evaluate how well the language design works in practice, we ported all 10 general-purpose implementations in the Java collections framework (JCF) as well as relevant interfaces and abstract implementations, to Genus. The result is a safer, more precise encoding and more code reuse with little extra programmer effort.

The single most interesting constrained generic in JCF is probably TreeSet (and TreeMap, which backs it). In its Java implementation, elements are ordered using either the element type’s implementation of Comparable or a comparator object passed as a constructor argument, depending on which constructor is used to create the set. This ad hoc choice results in error-prone client code. In Genus, by contrast, the ordering is part of the TreeSet type, eliminating 35 occurrences of ClassCastException in TreeSet’s and TreeMap’s specs.

Genus collection classes are also more faithful to the semantics of the abstractions. Unlike a Set[E], a List[E] should not necessarily be able to test the equality of its elements. In Genus, collection methods like contains and remove are instead parameterized by the definition of equality (§3.2). These methods cannot be called unless a model for Eq[E] is provided.

More powerful genericity also enables increased code reuse. For example, the NavigableMap interface allows extracting a descending view of the original map. In JCF, TreeMap implements this view by defining separate classes for each of the ascending and descending views. In contrast, Genus expresses both views concisely in a single class parameterized by a model that defines how to navigate the tree, eliminating 160 lines of code. This change is made possible by retroactive, non-unique modeling of compareTo().

Thanks to default models—in particular, implicit natural models, for popular operations including toString, equals, hashCode and compareTo—client and library code ordinarily type-check without using with clauses. When with clauses are used, extra expressive power is obtained. In fact, the descending views are the only place where with clauses are needed in the Genus collection classes.

8.2 Porting the Findbugs Graph Library to Genus

We ported to Genus the highly generic Findbugs [findbugs-release] graph library (∼1000 non-comment LoC), which provides graph algorithms used for the intermediate representation of static analyses. In Findbugs, the entities associated with the graph (e.g., Graph, Vertex, Edge) are represented as Java interfaces; F-bounded polymorphism is used to constrain parameters. As we saw earlier (§2), the resulting code is typically more cumbersome than the Genus version.

We quantified this effect by counting the number of parameter types, concrete types and keywords (extends, where) in each type declaration, ignoring modifiers and the name of the type. Across the library, Genus reduces annotation burden by 32% yet increases expressive power. The key is that constraints can be
Table 1: Comparing performance of Java and Genus

<table>
<thead>
<tr>
<th>data structure</th>
<th>Java (s)</th>
<th>Genus (s) [spec.]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Non-generic sort</td>
<td></td>
<td></td>
</tr>
<tr>
<td>double[]</td>
<td>1.3</td>
<td></td>
</tr>
<tr>
<td>Double[]</td>
<td>3.8</td>
<td></td>
</tr>
<tr>
<td>ArrayList[double]</td>
<td>—</td>
<td>5.4 [4.0]</td>
</tr>
<tr>
<td>ArrayList[Double]</td>
<td>9.6</td>
<td>14.5 [8.3]</td>
</tr>
<tr>
<td>Generic sort:</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Comparable[T]</td>
<td></td>
<td></td>
</tr>
<tr>
<td>double[]</td>
<td>—</td>
<td>19.3 [1.3]</td>
</tr>
<tr>
<td>Double[]</td>
<td>7.7</td>
<td>10.0 [3.8]</td>
</tr>
<tr>
<td>ArrayList[double]</td>
<td>—</td>
<td>6.7 [4.0]</td>
</tr>
<tr>
<td>ArrayList[Double]</td>
<td>9.8</td>
<td>17.9 [8.3]</td>
</tr>
<tr>
<td>ArrayLike[A,T], Comparable[T]</td>
<td></td>
<td></td>
</tr>
<tr>
<td>double[]</td>
<td>—</td>
<td>17.0 [1.3]</td>
</tr>
<tr>
<td>Double[]</td>
<td>12.8</td>
<td>12.4 [3.8]</td>
</tr>
<tr>
<td>ArrayList[double]</td>
<td>—</td>
<td>24.6 [4.0]</td>
</tr>
<tr>
<td>ArrayList[Double]</td>
<td>12.8</td>
<td>24.8 [8.3]</td>
</tr>
</tbody>
</table>

expressed directly without encoding them into subtyping and parametric polymorphism; further, prerequisite constraints avoid redundancy.

8.3 Performance

The current Genus implementation targets Java 5. To explore the overhead of this translation compared to similar Java code, we implemented a small Genus benchmark whose performance depends heavily on the efficiency of the underlying genericity mechanism, and hence probably exaggerates the performance impact of generics. The benchmark performs insertion sort over a large array or other ordered collection; the actual algorithm is the same in all cases, but different versions have different degrees of genericity with respect to the element type and even to the collection being sorted. Element type T is required to satisfy a constraint Comparable[T] and type A is required to satisfy a constraint ArrayLike[A,T], which requires A to act like an array of T’s. Both primitive values (double) and ordinary object types (Double) are sorted.

The results from sorting collections of 100k elements are summarized in Table 1. Results were collected using Java 7 on a MacBook Pro with a 2.6GHz Intel Core i7 processor. All measurements are the average of 10 runs, with an estimated relative error always within 2%. For comparison, the same (non-generic) algorithm takes 1.1s in C (with gcc -O3). The Java column leaves some entries blank because Java does not allow primitive type arguments.

To understand the performance improvement that is possible by specializing individual instantiations of generic code, we used hand translation; as mentioned above, the design of Genus makes such specialization easy to do. The expected performance improvement is shown in the bracketed table entries. Specialization to primitive types is particularly useful for avoiding the high cost of boxing and unboxing primitive values, but the measurements suggest use of primitive type arguments can improve performance even without specialization (e.g., Genus ArrayList[double] is usually faster than Java ArrayList<Double>).

9 Formalization and Decidability

This section formalizes the key aspects of the Genus type system. Since Genus includes models as part of types, the type checker would have trouble in interpreting types in signatures correctly when checking program well-formedness, if the program were fed to the type checker without default models being resolved in the first place. Therefore, we devote §9.2–9.4 to formalizing default model resolution, which works by translating one calculus into another—the source calculus allows default models while the target does not. §9.2 defines the semantic rules required by the translation, §9.3 specifies the translation procedure, and §9.4
proves its termination. §9.5 then gives the semantic rules for checking well-formedness in the translated calculus, and §9.6 defines the operational semantics. We are not aware of any unsoundness in the type system, but leave proving soundness to future work. We first present the syntax and the notation we use.

### 9.1 Syntax and notation

Figure 11 presents the syntax of the two calculi, \( \lambda^*_\text{Genus} \) and \( \lambda_{\text{Genus}} \). Default models are signified by \( * \) in \( \lambda^*_\text{Genus} \), while \( \lambda_{\text{Genus}} \) programs are \( *\)-free.

**Notation.** An overline denotes a (possibly empty) sequence. For example, \( \overline{X} \) denotes a sequence of type variables, and \( \overline{X}_i \) denotes the \( i \)-th element in that sequence. \( \emptyset \) denotes an empty sequence, and \( \#(\overline{X}) \) gives the size of \( \overline{X} \). We use the syntax \( \text{where } \overline{C}_1[\overline{\tau}_1] \circ_1; \overline{C}_2[\overline{\tau}_2] \circ_2 \) to distinguish between intrinsic constraints \( \overline{C}_1[\overline{\tau}_1] \circ_1 \) and extrinsic constraints \( \overline{C}_2[\overline{\tau}_2] \circ_2 \) in a where clause when necessary. It is implicitly required that the model variables of a where clause may not occur in the types of that where clause. To avoid clutter, we write \( [\beta] \) to denote (optional) universally or existentially quantified parameters, and write \( [\alpha] \) to denote the arguments to these optional parameters. We use \( \text{msig} \) and \( m : \tau \) interchangeably to represent method signatures. A model environment \( \Delta \) consists of model variables and (possibly universally quantified) models. It also records the constraint each model witnesses. We assume that class declarations, constraint declarations, model declarations, fields of a class and its superclasses, methods of a class, methods of a constraint and those of its prerequisite constraints, and variables of a method signature respectively have distinct names. Type variable names and model variable names are unique in any context. It is also assumed that class hierarchies and constraint hierarchies are well-founded—they do not contain cycles. The “=”
sign, when applied to two non-existential types, means syntactic equality. Functions $FTV(\cdot)$ and $FMV(\cdot)$ respectively return free type variables and free model variables of a syntactic object.

**Simplifications.** The calculi are in the style of Featherweight Java (FJ) [Igarashi et al. 2001], and hence those simplifications akin to FJ. To concentrate on formalizing Genus-style type constraints, we omit subtyping constraints. We also omit a few Genus-specific features that add to the convenience of the mechanism but have little impact on expressivity: a where-clause constraint is always associated with a model variable; method calls are always explicitly qualified with expanders, and method declarations are always explicitly qualified with receiver types; and reusing models via model inheritance or model projection are not supported. Features requiring load-time checking are not captured. We consider that type arguments and models for intrinsic constraints are given for method calls, as prior formalizations in the FJ family do. Nevertheless, models for extrinsic constraints may be omitted (i.e., denoted as $*$), forcing the compiler to resolve default models.

The formalization handles existentials differently from Genus. Receivers of field accesses and method calls are restricted to expressions of non-existential types; existential types can only be used as method return types. This treatment forces existentially quantified type parameters and where-clause constraints in $\lambda_{Genus}$ to always be explicitly captured by universally quantified ones. Meanwhile, because of existential types, in method calls, elided type arguments and models for intrinsic constraints—if they can be inferred via unification—cannot always be written in the calculi. We focus on the semantics of models in this work, while leaving fully modeling existentials in Genus to future work. One possibility is to explore how previous approaches to formalizing Java wildcards (e.g., Cameron et al. [2008]) fit in.

**Substitution.** Substitution takes the form $\cdot\{\cdot/\cdot\}$, and is defined in the usual way. It is required that the item that is substituted into, and the item that is used to substitute, be $*$-free. Substitutions succeed only when the size of the argument list and that of the formal parameter list match. For example, $\cdot\{\tau/X\}$ substitutes $\tau^i$ for $X^i$ in the first argument, with $\#(\tau) = \#(X)$ assumed. We introduce an abbreviated notation for substituting types and models at once: $\cdot\{\alpha/\beta\}$ is shorthand for $\cdot\{\pi/\sigma\}\{\tau/X\}$ where $[\alpha]$ is $[\tau$ with $\pi]$ and $[\beta]$ is $[X$ where $C[\tau]' o]$.

### 9.2 Static semantics (Part I)

This section defines the semantic rules required by the translation from $\lambda^*_{Genus}$ to $\lambda_{Genus}$ (i.e., resolving default models in $\lambda^*_{Genus}$ programs).

**Variance.** Figure 12 defines the $\text{vary}^\rho(i, C)$ relation that says the $i$-th type parameter of constraint declaration $C$ is contravariant or covariant when $\rho$ is $-$ or $+$ respectively. It is possible for $\text{vary}^- (i, C)$ and $\text{vary}^+ (i, C)$ to hold simultaneously, in which case $C$ is $bivariant$ with respect to the type parameter. The relation is defined using three other relations, $\text{vary}^\rho (X, C[\tau])$, $\text{vary}^\rho (X, \tau)$, and $\text{vary}^\rho (X, \dot{\tau})$. They respectively describe the variance of a type variable in a constraint, a method signature, and a type. Notice that these rules apply whether the constraint declaration contains default models or not, because $*$ can only occur in invariant positions.

**Subtyping and constraint entailment.** Figure 13 defines the subtyping relation $\leq$ used by the translation. They are used in checking constraint entailment relationships (Figure 14). Rules are only defined for non-existential types because checking constraint entailment does not require checking subtyping relationships that involve existential types. They are enriched by subtyping rules defined in Figure 23—those rules deal with existential types—in order to be used for type-checking $\lambda_{Genus}$ expressions.
\[
\begin{align*}
\text{Figure 12: Variance relations. } \rho \text{ ranges over } \{+,-\}, \text{ and } -\rho \text{ reverses the sign denoted by } \rho.
\end{align*}
\]
Figure 14 defines the constraint-entailment relation $\leq$. Rule (E-PREREQ) shows that a constraint entails its prerequisite constraints, while rules (E-) and (E+) show that given the variance of a type parameter, constraint entailment reduces to subtyping. Invariance is covered by rule (E-REFL), and bivariance is down-graded to contravariance.

A requirement established in §9.4 guarantees that either side of $\leq$—and hence $\leq$—does not contain default models. In §9.3 and §9.4, constraint entailment relationships derived without using constraint covariance (i.e., rule (E+)) are denoted by $\leq_-$.

### 9.3 Translating $\lambda_{\text{Genus}}$ to $\lambda_{\text{Genus}}$

The translation from $\lambda_{\text{Genus}}^*$ to $\lambda_{\text{Genus}}$ is defined with the help of translation contexts. A translation context takes the form

$$[\mathcal{P}]^{\Delta}_{\tau} [:]$$

in which $\mathcal{P}$ denotes a syntactic category in $\lambda_{\text{Genus}}^*$, $\Delta$ is the model environment consisting of enabled models, $\tau$ records additional information required to translate a term of $\mathcal{P}$ ($\tau$ may or may not be present depending on $\mathcal{P}$), and $[\cdot]$ is the hole for which a subterm of $\mathcal{P}$ is to be substituted.

Figure 15 defines all translation contexts. Most of the translation contexts are mundane: they simply propagate enabled models into model environments that are used to translate subterms. For example, in translating a class declaration, the models enabled by the where clause of that class declaration are added to the model environment when translating its superclass, field declarations, and method declarations, as shown by the definition of $[\text{idecl}]^{\Delta}$. The definition in Figure 15 uses the auxiliary function $\beta(\cdot)$ to access the parameters of a generic, and the auxiliary function models$(\cdot)$ to obtain the models enabled by a where clause or the use declarations of a program.

Some of the terms in Figure 15 are highlighted, indicating that they may contain default models. So they are the places where translation is going to happen next. Also, the notation for sequences is abused so that $[\mathcal{P}]^{\Delta}_{\tau}$ means $\mathcal{P}, [\mathcal{P}]^{\Delta}_{\tau}$. Notice that the types $\mathcal{P}$ in a constraint $\mathcal{C}[\mathcal{P}]$ are never highlighted. The reason is given in §9.4.

Figure 16 defines the relation $\mathcal{P} \xrightarrow{\Delta}_{\tau} \mathcal{P}'$, which reads “the $\lambda_{\text{Genus}}$ term $\mathcal{P}$ translates to the $\lambda_{\text{Genus}}$ term $\mathcal{P}'$ under model environment $\Delta$ with additional information $\tau$.” Every translation context (except $[\cdot]^{\Delta}_{\mathcal{C}[\mathcal{P}]}$) represents a context rule. Nevertheless, the goal of the translation is to replace $\ast$ with resolved models. So $[\ast]^{\Delta}_{\mathcal{C}[\mathcal{P}]}$ is defined as a hole (i.e., $[\cdot]^{\Delta}_{\mathcal{C}[\mathcal{P}]}$), and the rewriting rule it corresponds to is where the actual translation takes place. Translating $\ast$ needs additional information, namely the constraint for which the default model is to be resolved, and hence the bottom-right corners of $[\alpha]^{\Delta}_{\beta}$ and $[\ast]^{\Delta}_{\mathcal{C}[\mathcal{P}]}$.

The rewriting rule depends on the relation $\Delta \vdash \mathcal{C}[\mathcal{P}] \models \mu$, defined in Figure 17 by rule (D), to obtain the resolved model. Rule (D) states that if there is exactly one model that is enabled for constraint $\mathcal{C}[\mathcal{P}]$ in the model environment $\Delta$, the default model for that constraint is resolved to that unique witness. Function enabled$^{\Delta}(\mathcal{C}[\mathcal{P}])$, defined extensionally, then tells if a model is enabled in the environment for the given constraint. In order to focus on constraint variance and recursive resolution, which are more interesting because of their implications for decidability, $\lambda_{\text{Genus}}^*$ does not enable natural models or uniquely satisfying models. Apart from these simplifications, the formalization matches the description in §§4.4, §4.7 and §5.2: the model variable in (D-MVAR) comes from a where-clause constraint, the model in (D-MODEL) comes from a use declaration or a model declaration, and (D-E) allows using constraint contravariance to derive enabled models.

In rule (D-MODEL), $\overline{\tau}$ and $\overline{\mu}$ are considered to be inferred via unification. Their well-formedness is not enforced at this stage, but is instead checked during type checking. Notice that the premise of (D-MODEL) contains recursive applications of (D). It is for resolving default models for extrinsic constraints of use declarations. The use of $\leq_-$ instead of $\leq$ in rule (D-E) is justified in §9.4.
\[ \text{prog}^\beta \quad ::= \quad \langle \text{use}^\beta, \text{idc}l^\beta, \text{odec}l^\beta, \varepsilon \rangle \mid \langle \text{use}, \text{odec}l^\beta \rangle \mid \langle \text{use}, \text{odec}l^\beta \rangle \mid \langle \text{use}, \text{odec}l^\beta \rangle \mid \langle \text{use}, \text{idc}l^\beta \rangle \mid \langle \text{odec}l^\beta \rangle \mid \langle \text{odec}l^\beta \rangle \mid \langle \text{odec}l^\beta \rangle \\
\]

\[ \text{use} \quad ::= \quad [\beta] \text{use } M \left[ a \left[ \Delta_{\beta(M)} \right] \right] \quad \text{for } C[\tau] \]

\[ \text{idc}l^\Delta \quad ::= \quad \text{class } L_1 [\beta] \text{ extends } L_2 \left[ a \left[ \Delta_{\beta(L_2)} \right] \right] \left\{ \text{idc}l^\Delta, \text{mdecl} \right\} \mid \text{class } L_1 [\beta] \text{ extends } L_2 [\alpha] \left\{ \text{idc}l^\Delta, \text{mdecl} \right\} \]

\[ \text{odec}l^\Delta \quad ::= \quad \text{model } M \left[ X \right] \text{ where } C_1 [\tau] \quad \text{for } C_2 [\tau] \left\{ \text{mdecl} \right\} \left\{ \text{models} \right\} \]

\[ \text{fdecl}^\Delta \quad ::= \quad \langle \tau \rangle^\Delta \text{ f} \]

\[ \text{mdecl}^\Delta \quad ::= \quad \text{msig}^\Delta \left\{ \langle \varepsilon \rangle \mid m : \tau \left\{ e \right\}^\Delta \text{models} \left( \beta(m) \right) \right\} \]

\[ \text{msig}^\Delta \quad ::= \quad \langle \tau \rangle^\Delta \text{ models} \left( \beta \right) \mid \langle \tau \rangle^\Delta \text{ m} \mid \langle \tau \rangle^\Delta \mid \langle \tau \rangle^\Delta \text{ models} \left( \beta \right) \]

\[ \langle \tau \rangle^\Delta \quad ::= \quad \Delta_{\beta(L)}^\Delta \]

\[ \text{models} \left( \left[ X \right] \right) = \text{o} \quad \text{models} \left( \left[ C[\tau] \right] \right) = \beta \quad \text{models} \left( \left[ C[\tau] \right] \right) = \beta \quad \text{models} \left( \left[ C[\tau] \right] \right) = \beta \]

\[ \text{locl}^\Delta \quad ::= \quad \text{e}^\Delta \text{. } \text{f} \mid \text{new } L[\alpha]^\Delta \left( f = \varepsilon \right) \mid \text{new } L[\alpha]^\Delta \left( f = \varepsilon^\Delta \right) \]

\[ \text{e}^\Delta \quad ::= \quad e \left( \langle \tau \rangle^\Delta \text{. } m \right) \left[ \alpha \right] \left( \varepsilon \right) \mid e \left( \tau \cdot m \right) \left[ \alpha \right] \left( \varepsilon^\Delta \right) \mid e \left( \tau \cdot m \right) \left[ \alpha \right] \left[ \varepsilon \right] \mid e \left( \mu \cdot m \right) \left[ \alpha \right] \left( \varepsilon^\Delta \right) \mid e \left( \mu \cdot m \right) \left[ \alpha \right] \left[ \varepsilon \right] \]

\[ \text{Figure 15: Translation contexts} \]

\[ \text{prog}_1^{\Delta_1} \quad ::= \quad \ldots \text{prog}_2^{\Delta_2} \quad \ldots \quad \text{prog}_1 \neq * \]

\[ \text{context rules} \]

\[ \text{locl}^\Delta \quad ::= \quad \text{e}^\Delta \quad \text{f} \mid \text{new } L[\alpha]^\Delta \left( f = \varepsilon \right) \mid \text{new } L[\alpha]^\Delta \left( f = \varepsilon^\Delta \right) \]

\[ \text{Figure 16: Translating } \lambda_{Genus}^x \text{ into } \lambda_{Genus} \]

22
9.4 Termination of default model resolution

We first illustrate the termination conditions for default model resolution. First, recursive resolution in the presence of recursive subclassing may result in nontermination. Consider the λGenus program in Figure 18, in which \(L_2\) subclasses \(L_1\) in a recursive way. Suppose \(C\) is contravariant with respect to its type parameter. To search for a model for constraint \(C[L_2]\), rule (D-E) is applied since \(C[L_1[L_2]] \not\leq C[L_2]\). Consequently, a model for \(C[L_1[L_2]]\) is wanted. We are then directed to examine the default model candidacy of the use declaration for \(C[L_1[L_2]]\): the type parameter \(X\) is inferred as \(L_2\), and a model for the extrinsic constraint \(C[X]\{L_2/X\}\) becomes the new goal, causing the search to cycle.

Classes like \(L_2\) are called “shapes” in Greenman et al. [2014]. In practice, “shapes” encode methods like `compareTo` and `compareTo`. Since these methods are effectively encoded by constraints in Genus, Genus disallows recursive subclassing and F-bounds, effectively enforcing Material–Shape Separation to prevent nontermination. The first termination condition follows:

**Requirement 1.** No recursive subclassing in \(\lambda_{Genus}\).

The premise of (D-E) contains a \(\not\leq\) judgment (recall that \(\not\leq\) means the constraint entailment relationship is derived without using constraint covariance). Similarly to how the coupling of constraint contravariance and recursive subclassing can cause default model resolution not to terminate, the coupling of constraint covariance and recursive superclassing can too. We choose to simply refrain from using constraint covariance in searching the model environment, because constraint covariance is much less common than constraint contravariance, and because recursive superclassing cannot be avoided (consider the subtyping relationship between `List[Object]` and `Object`).

Second, we adapt the “Paterson conditions” [Sulzmann et al. 2007] studied in the Haskell setting:

**Requirement 2.** Denote a “use” declaration by \(\text{use } [X \text{ where } \alpha_1; \alpha_2, \ldots] \mathcal{M} \alpha \text{ for } C[X]\).
\[ \|\tau\| = \min_i \|\tau^i\| \]
\[ \|X\| = 1 \]
\[ \|L[\alpha]\| = 1 + \langle L[\alpha]\rangle \]
\[ \|\tau \text{ with } \overline{\tau}\| = \|\tau\| \]
\[ \langle L[\alpha]\rangle = \max \left\{ \|\tau\|, \max_{L[\alpha] \leq L'[\alpha']} \langle L'[\alpha']\rangle \right\}. \]

**Figure 19:** The ranking function \(\|\cdot\|\) used in Requirement 2

...  
constraint \(C[\tau] \{\ldots\}\)  
class \(L_1[\tau \text{ with } *] \{\ldots\}\)  
class \(L_2[\tau \text{ with } *] \{\ldots\}\)  
...

**Figure 20:** The presence of default models in constrained types leads to a scheduling deadlock.

- **No type variable of a use declaration has more occurrences in any extrinsic constraint \(C_2[\overline{\tau_2}]\) than \(C_3[\overline{\tau_3}]\).**
- **For all \(i\), \(\|\overline{\tau_2}\| < \|\overline{\tau_3}\|\), where \(\|\cdot\|\) is the ranking function defined in Figure 19.**

The requirement is motivated by reducing the rank of the constraint to resolve by at least one each time rule (D-MODEL) is applied. The definition of \(\|L[\alpha]\|\) takes subclassing into consideration, ensuring \(\|L[\alpha]\| \geq \|L'[\alpha']\|\) if \(L[\alpha]\) is a subclass of \(L'[\alpha']\). It is implied that models have zero rank, so unlike the occurrence of a class name in its superclasses’ type arguments being banned by Requirement 1, the occurrence of a class name in its superclasses’ with clauses is not a concern. Requirement 2 prevents use declarations like “use DualGraph;” from §4.7.

The third condition has to do with the scheduling of resolution. Consider the \(\lambda^{*}\text{Genus}\) program in Figure 20. Resolving the first occurrence of \(*\) requires knowing the type \(L_1[\tau \text{ with } *]\) in \(L_2\)'s where clause, which contains the second occurrence of \(*\). Resolving the second occurrence of \(*\) then requires knowing the type \(L_2[\tau \text{ with } *]\) in \(L_1\)'s where clause, which contains the first occurrence of \(*\), leading to circular dependency. The issue is addressed by requiring constrained types (i.e., the types \(\overline{\tau}\) in a constraint \(C[\overline{\tau}]\)) be free of default models:

**Requirement 3.** In \(\lambda^{*}\text{Genus}\), symbol \(*\) does not appear in constrained types.

In addition to ruling out a source of nontermination, the requirement saves programmers from the arguably hazardous indirectness in reasoning about where-clause constraints, by forcing constrained types to be expressed clearly.

Now we are ready to prove that default model resolution in \(\lambda^{*}\text{Genus}\) terminates. The termination of the underlying unification algorithm is reasonably assumed. We sketch in Figure 21 the procedure that implements enabled\(\Delta(C[\overline{\tau}])\) since the algorithm for deciding default model candidacy is not very straightforward from the inference rules in Figure 17. Termination is proved with respect to the algorithm.

**Lemma 1.** All subtyping derivations are finite in \(\lambda^{*}\text{Genus}\).
Proof. By the well-foundedness of class inheritance.

Corollary 1. All constraint entailment derivations are finite in $\lambda^*_\text{Genus}$.

Proof. By Lemma 1.

Theorem 1. Default model resolution terminates in $\lambda^*_\text{Genus}$.

Proof. By Corollary 1, the use of constraint entailment in the algorithm is not a concern. The algorithm has two parts. The first part investigates the default model candidacy of model variables. This part, corresponding to (D-MVAR) and (D-E), terminates trivially. The second part of the algorithm investigates the default model candidacy of possibly universally quantified models. This part, corresponding to (D-MODEL) and (D-E), obtains the substitutions (i.e., $\Delta$), and by Lemma 1.

Figure 21: Algorithm that searches the model environment $\Delta$ for enabled models for constraint $C[\tau]$.

Proof. By Corollary 1.

Theorem 1. Default model resolution terminates in $\lambda^*_\text{Genus}$.

Proof. By Corollary 1, the use of constraint entailment in the algorithm is not a concern. The algorithm has two parts. The first part investigates the default model candidacy of model variables. This part, corresponding to (D-MVAR) and (D-E), terminates trivially. The second part of the algorithm investigates the default model candidacy of possibly universally quantified models. This part, corresponding to (D-MODEL) and (D-E), obtains the substitutions (i.e., $\tau_1$ and $\tau_2$) for the quantified type parameters and intrinsic constraints through procedure $\text{unify}$, and triggers default model resolution recursively for extrinsic constraints $C_2[\tau_2]\{\tau_1/\tau\}$ (i.e., $\tau_3$). Hence, the termination of this part depends on the well-foundedness of the recursion, given the assumption that unification terminates. Define a ranking function $|\cdot|$ on sequences of types to represent the depth of the recursion:

$$|\tau| = 1 + \max_{\langle X | C_1[\tau_1] \alpha_1; C_2[\tau_2] \alpha_2; M[\alpha] : C[\tau_3] \in \Delta \text{ such that } C_2[\tau_2]\{\tau_1/\tau\}\{\tau_3/\tau\} \preceq C[\tau]}} \max_i \left| \frac{\tau_2}{\tau_1} \{\tau_4/\tau\} \right|.$$

It remains to show that the ranking function is well-defined. In fact, using the termination conditions, we can show $\left| \frac{\tau_2}{\tau_1} \{\tau_4/\tau\} \right| < |\tau|$ for all $i$, from which the well-definedness of $|\cdot|$ follows.

First of all, the ranking function $|\cdot|$ is well-defined by structural induction and by Requirement 1.

Now, by the second part of Requirement 2, we know $\left| \frac{\tau_2}{\tau_1} \right| < |\tau_2|$. And by the first part of Requirement 2, we then have

$$\left| \frac{\tau_2}{\tau_1} \{\tau_4/\tau\} \right| < |\tau_2\{\mu_1/\sigma_1\}\{\tau_4/\tau\}|.$$  (1)
Meanwhile, by the unification constraint $C_3[\tau_3]\{\mu_3/\overline{\alpha_1}\}\{\tau_3/X\} \preceq C[\tau]$, each $\tau^j$ must be either identical to or a subtype of $\tau^k\{\mu_1/\overline{\alpha_1}\}\{\tau_1/X\}$ for some $k$, which gives us
\[
\|\tau^j\| \geq \|\tau^k\{\mu_1/\overline{\alpha_1}\}\{\tau_1/X\}\| \geq \|\tau_3\{\mu_1/\overline{\alpha_1}\}\{\tau_4/X\}\|
\]
by the definition of $\|\cdot\|$. Since the above inequality holds for all $j$,
\[
\|\tau\| = \min_j \|\tau^j\| \geq \|\tau_3\{\mu_1/\overline{\alpha_1}\}\{\tau_4/X\}\|.
\]
(2)

The result follows from (1) and (2). \qed

9.5 Static semantics (Part II)

This section defines the rest of the static semantics in terms of $\lambda_{\text{Genus}}$, in which all models are explicit.

Lookup functions. Figure 22 defines the auxiliary lookup functions. Function fields ($L[\alpha]$) gives all the field members defined in class $L[\alpha]$ and its superclasses. Function mtype ($m, L[\alpha]$) returns the signature of the most specific method with name $m$ defined in class $L[\alpha]$ or its superclasses, while function mtype ($m, C[\tau]$) returns the signature of a method with name $m$ defined in constraint $C[\tau]$ or its prerequisite constraints.

Subtyping. The subtyping rules defined in Figure 23 enriches those in Figure 13 to handle existential types.

In (S), \$t$ and \$t_4$ are considered to be generated via unification to replace $X^j_{\tau_4}$ and $\overline{\alpha}_4$, with the unification constraint being $\emptyset; \emptyset \vdash \tau_3 \leq \tau_3\{\mu/\overline{\alpha}\}\{\tau_4/X\}$. $X^j_{\tau_4}$ and $\overline{\alpha}_4$ denote the existentially quantified type parameters and intrinsic constraints of the supertype. We assume the unification algorithm to be complete but not necessarily sound, meaning that it produces the correct solution whenever the unification constraints are solvable, although it may produce a bogus one when the constraints are unsolvable. Invalid solutions are rejected by well-formedness checks imposed on \$t$ and \$t_4$.

Rule (S) also requires that any extrinsic constraints of the supertype be entailed by its positionally corresponding extrinsic constraint in the subtype. Positional correspondence eases reasoning about subtyping and joins of types.

Subtyping transitivity and reflexivity in the presence of existential types are results of (S).
well-formedness is formalized using the constraints of the existential type on the right-hand side of $\leq$ (i.e., default model resolution). Rule (C-D) shows that default models should be resolved for the extrinsic coercions admitting subtyping. Coercion may also induce extra context-dependent computation (i.e., default model resolution). Rule (C-D) shows that default models should be resolved for the extrinsic coercions admitting subtyping.

Well-formedness rules are decomposed into rules for type well-formedness, constraint well-formedness (Figure 24), model well-formedness (Figure 25), expression well-formedness (Figure 26), and program well-formedness (Figure 29).

Similarly to expression well-formedness being formalized into the typing relation $\Gamma; \Delta; E \vdash e : \tau$, model well-formedness is formalized using the witnessing relation $\Gamma; \Delta; \mu : C[\tau]$ indicating that the model $\mu$ witnesses the constraint $C[\tau]$ in the given context.

Expression typing rules (Figure 26) are mostly standard except that expanders and with clauses play a part in method calls. Rules (W-TCALL) and (W-MCALL) are for type-checking method calls whose expanders are types and models respectively.
\[ \Gamma; \Delta \vdash \mu :: C[\tau] \]

\[
\text{model } M \mid X \text{ where } C_1[\tau_1] \circ C_2[\tau_2] \{ \ldots \}
\]

\[
(\forall i) \Gamma; \Delta \vdash \tau_i \vdash \text{ok} \quad (\forall i) \Gamma; \Delta \vdash \tau_i :: C_1[\tau_1] \{ \tau_i / X \}
\]

\[
\Gamma; \Delta \vdash M[\tau] \text{ with } \pi :: C_2[\tau_2] \{ \pi / \sigma \} \{ \tau_3 / X \}
\]

\[
o :: C[\tau] \in \Delta
\]

\[ (W-MVAR) \quad \Gamma; \Delta \vdash o :: C[\tau] \]

\[ (W-M)
\]

\[ (E-SUBSUME) \quad \Gamma; \Delta \vdash \mu :: C_1[\tau_1] \quad \vdash C_1[\tau_1] \leq C_2[\tau_2] \]

Figure 25: The witnessing relation

The helper relations in Figure 27 are for checking well-formedness of class declarations and constraint declarations: relations receiver-ok (\( \tau, L \)) and receiver-ok (\( \tau, C \)) ensures that method receiver types in class declarations and constraint declarations are as expected, and relation \( \Gamma; \Delta \vdash \text{override-ok} (m, L[\alpha], \tau) \) indicates that class method \( m : \tau \) has a signature that appropriately overrides the method with the same name available in the superclass \( L[\alpha] \).

Figure 28 defines the vocabulary to be used in checking model–constraint conformance. Relation \( \Gamma; \Delta \vdash \text{specialize-ok} (m, C[\tau], \tau) \) indicates that method \( m : \tau \) (coming from some model) has a signature that appropriately specializes the method with name \( m \) required by constraint \( C[\tau] \). The model method can only specialize the receiver type and those argument types whose counterparts in the constraint method are a type parameter of the constraint declaration. Relation \( \Gamma; \Delta \vdash \text{model-ok} \left( \text{msig}, C[\tau] \right) \) indicates that \( \text{msig} \), the methods of some model, models constraint \( C[\tau] \). Specifically, it demands that for all methods required by the constraint there be a unique most specific method in \( \text{msig} \) for every possible combination of dynamic types of the method arguments and the receiver allowed by the required method. Therefore, four other relations are needed:

- Relation \( \text{applicable}^{\Gamma; \Delta} \left( \tau, L_1[\alpha_1], L_2[\alpha_2] \right) \) indicates that \( L_1[\alpha_1] \) and \( L_2[\alpha_2] \), representing the types of the receiver and the arguments at run time, are applicable to the method signature \( \tau \).
- Relation \( \text{noLessSpecific} (\tau_1, \tau_2) \) indicates that \( \tau_1 \) is a method signature that is at least as specific as \( \tau_2 \).
- Relation \( \Gamma; \Delta \vdash \text{subordinate} (\tau_1, \tau_2, L_1[\alpha_1], L_2[\alpha_2]) \) indicates that method signature \( \tau_1 \) is subordinate to \( \tau_2 \) with respect to the set of dynamic types, meaning either \( \tau_1 \) is not applicable to the dynamic types, or it is applicable but is less specific than \( \tau_2 \).
- Relation \( \Gamma; \Delta \vdash \text{mostSpecific} (j, \tau, L_1[\alpha_1], L_2[\alpha_2]) \) indicates that \( \tau^j \) is the most specific method signature among \( \tau \) for the set of dynamic types.

Rule (W-ODECL) shows that checking well-formedness of model declarations uses both specialize-ok and model-ok.

### 9.6 Operational semantics

Figure 30 presents the values and evaluation contexts in \( \lambda_{\text{Genus}} \). The auxiliary lookup functions defined in Figure 31 find the most specific method definition for method calls: function mdecl \( (m, L[\alpha]) \) finds the most specific method definition with name \( m \) defined by class \( L[\alpha] \) and its superclasses, and function mdecl \( (m, M[\alpha_0], L_1[\alpha_1], L_2[\alpha_2]) \) finds for a receiver object of type \( L_1[\alpha_1] \) and arguments of types \( L_2[\alpha_2] \) the most specific method definition with name \( m \) defined by model \( M[\alpha_0] \). Figure 32 presents the reduction rules.
\[ \Gamma; \Delta; E \vdash \dot{e} : \dot{\tau} \]

\[ \Gamma; \Delta; E \vdash \dot{e} : \dot{\tau}_1 \leq \dot{\tau}_2 \quad \text{(S-SUBSUME)} \]

\[ \Gamma; \Delta; E \vdash x : E(x) \quad \text{(W-VAR)} \]

\[ \text{class } L [\beta] \ldots \{ \tau \} \text{ undeclared} \]

\[ \Gamma; \Delta; E \vdash e : L[\alpha] \quad \text{(W-FIELD)} \]

\[ \Gamma; \Delta; E \vdash e.\dot{f} : \tau^i{\alpha/\beta} \quad \text{(W-TCALL)} \]

\[ \text{mtype} \ (m, L[\alpha]) = [\dot{X} \text{ where } \dot{\tau}_1 \dot{\alpha}_1; \dot{\tau}_2 \dot{\alpha}_2 (L[\alpha], \dot{\tau}_3) \rightarrow \dot{\tau}_4] \]

\[ \text{mtype} \ (m, C[\tau_3]) = [\dot{X} \text{ where } \dot{\tau}_1 \dot{\tau}_2 \dot{\alpha}_1; \dot{\tau}_2 \dot{\alpha}_2 (\dot{\tau}_3, \dot{\tau}_4) \rightarrow \dot{\tau}_5] \]

\[ \Gamma; \Delta; E \vdash e.\dot{m}_L \dot{\rightarrow} \dot{\tau}_4 \quad \text{(W-MCALL)} \]

\[ \Gamma; \Delta; E \vdash e.\dot{m}_L \dot{\rightarrow} \dot{\tau}_5 \quad \text{(W-NEW)} \]

\[ \Gamma; \Delta; E \vdash e.\dot{m}_L \dot{\rightarrow} \dot{\tau}_1 \quad \text{(C-SUBSUME)} \]

\[ \text{receiver-ok} (\tau, L) \quad \text{receiver-ok} (\tau, C) \]

\[ \text{class } L [\dot{X} \text{ where } \dot{C} [\tau]] \ldots \{ \ldots \} \quad \text{constraint } C [\dot{X}] \ldots \{ \ldots \} \]

\[ \tau_1 = L[\dot{X} \text{ with } \dot{\beta}] \quad \text{(W-RECV-L)} \]

\[ \text{receiver-ok} ([\beta] (\tau_1, \tau_2) \rightarrow \tau_3, L) \quad \text{(W-RECV-C)} \]

\[ \text{mtype} \ (m, L[\alpha]) \text{ is undefined} \quad \text{(W-OVER-UNDEF)} \]

Figure 26: Expression typing

Figure 27: Helper relations for checking well-formedness of class and constraint declarations
Figure 28: Helper relations for checking well-formedness of model declarations
\[
\begin{align*}
\Gamma; \Delta \vdash \text{msig ok} \\
(\forall i) \overline{\alpha_i} \in \text{FMV}(\tau_3) \quad (\forall i) \overline{\alpha_i} \notin \text{FMV}(\tau_3) \\
\Gamma, X; \Delta \vdash C[\tau_1] \quad \Gamma; \Delta \vdash C_2[\tau_2] \quad \Gamma, X; \Delta, o : C_1[\tau_1] \vdash \tau_3 \quad \Gamma; \Delta \vdash \tau_4 \quad \Gamma, X; \Delta, o : C_1[\tau_1] \vdash \tau_5 \quad (W-\text{MSIG})
\end{align*}
\]
\[
\begin{align*}
\Gamma; \Delta \vdash \text{decl ok} \\
\Gamma; \Delta \vdash \text{decl ok} \\
\Gamma, X; \Delta, o : C[\tau_3]; \text{this} : \tau_2, x : \tau_1 \vdash \varepsilon ; \tau_1 \quad \Gamma, X, \Delta, o : C[\tau_3] \vdash \tau_1 \leq \varepsilon \quad (W-\text{MDECL})
\end{align*}
\]
\[
\begin{align*}
\Delta \vdash \text{lodecl ok} \\
\Delta' = \Delta, o : C[\tau_1] \\
\Delta' \vdash L_2[\alpha] \quad (\forall i) X; \Delta' \vdash \tau_i \quad (\forall i) X; \Delta' \vdash m : \tau \{\varepsilon\} \quad (W-\text{LDECL})
\end{align*}
\]
\[
\begin{align*}
\Delta \vdash \text{decl ok} \\
\Delta \vdash \text{decl ok} \\
(\forall i) X; \Delta \vdash \overline{\tau_1} \quad (\forall i) X; \Delta \vdash \overline{\tau_2} \quad (\forall i) \text{receiver-ok} \left(\overline{\tau_i}, L_1\right) \\
\Delta \vdash \text{constraint} C_1 \left\{X_1\right\} \text{extends} C_2 \left\{X_2\right\} \left\{m : \tau\right\} \quad (W-\text{CDECL})
\end{align*}
\]
\[
\begin{align*}
\Delta' = \Delta, o : C[\tau_1] \\
\Delta' \vdash M[X \text{ with } \overline{\tau}] : C_2[\tau_2] \vdash m : \tau \{\varepsilon\} \quad (\forall i) X; \Delta' \vdash \text{model-ok} \left(m : \tau, C_2[\tau_2]\right) \\
\Delta \vdash \text{specialize-ok} \left(m : \tau, C_2[\tau_2], \tau_i\right) \\
\Delta \vdash \text{model} M \left\{X \text{ where } C_1[\tau_1] \right\} \text{ for } C_2[\tau_2] \left\{m : \tau \{\varepsilon\}\right\} \quad (W-\text{ODECL})
\end{align*}
\]
\[
\begin{align*}
\Delta \vdash \text{use ok} \\
(\forall i) X_i \in \text{FTV}(\tau_3) \quad (\forall i) \overline{\alpha_i} \in \text{FMV}(\tau_3) \quad (\forall i) \overline{\alpha_i} \notin \text{FMV}(\tau_3) \\
(\forall i) X; \emptyset \vdash C[\tau_1] \quad (\forall i) X; \emptyset \vdash C_2[\tau_2] \quad X; o_1 : C_1[\tau_1] \vdash C_3[\tau_3] \quad X; o_2 : C_2[\tau_2] \vdash M[\alpha] : C_3[\tau_3] \\
\vdash \left\{X \text{ where } C_1[\tau_1] \right\} \text{ for } C_2[\tau_2] \left\{o_1 : C_2[\tau_2], o_2 : C_3[\tau_3]\right\} \quad (W-\text{USE})
\end{align*}
\]
\[
\begin{align*}
\vdash \left\{\beta \text{ use } M[\alpha] \text{ for } C[\tau_1] \right\} \text{ ok} \quad \Delta = [\beta] M[\alpha] : C[\tau] \\
(\forall i) \Delta \vdash \text{decl ok} \quad (\forall i) \Delta \vdash \text{decl ok} \quad (\forall i) \Delta \vdash \text{decl ok} \quad \emptyset; \Delta; \emptyset \vdash \varepsilon : \tau \text{ for some } \tau \\
\vdash \left\{\beta \text{ use } M[\alpha] \text{ for } C[\tau], \text{decl, decl, decl, } \varepsilon \right\} \text{ ok} \quad (W-\text{PROG})
\end{align*}
\]

Figure 29: Program well-formedness
models are chosen modularly at compile time. Not currently implemented but could be fully supported by Genus along with separate compilation, because dynamic compilation makes compilation time depend on the amount of generic library code. The OO language plates are not checked until instantiation, so developers see confusing error messages, and the lack of retroactive modeling [Austern 1998]. This pattern is used extensively in the STL and Boost libraries. Temporal focus here on more recent work that follows Haskell’s type classes in supporting retroactive modeling, complementing the discussion in previous sections.

Much prior work on parametric genericity mechanisms (e.g., [Liskov et al. 1984; Schaffert et al. 1986; Chambers 1992; Myers et al. 1997; Agesen et al. 1997; Cartwright and Steele Jr. 1998; Bracha et al. 1998; Kennedy and Syme 2001]) relies on constraint mechanisms that do not support retroactive modeling. We focus here on more recent work that follows Haskell’s type classes in supporting retroactive modeling, complementing the discussion in previous sections.

The C++ community developed the Concept design pattern, based on templates, as a way to achieve retroactive modeling [Austern 1998]. This pattern is used extensively in the STL and Boost libraries. Templates are not checked until instantiation, so developers see confusing error messages, and the lack of separate compilation makes compilation time depend on the amount of generic library code. The OO language G [Siek and Lumsdaine 2011], based on System $F^G$ [Siek and Lumsdaine 2005], supports separate compilation but limits the power of concept-based overloading. By contrast, C++ Concepts [Gregor et al. 2006] abandon separate compilation to fully support concept-based overloading. It was not adopted by the C++11 standard [Siek 2011], however. Concept-based overloading is orthogonal to the other Genus features; it is not currently implemented but could be fully supported by Genus along with separate compilation, because models are chosen modularly at compile time.
In Scala, genericity is achieved with the Concept design pattern and implicits [Oliveira et al. 2010]. This approach is expressive enough to encode advanced features including associated types [Chakravarty et al. 2005] and generalized constraints [Emir et al. 2006]. Implicits make using generics less heavyweight, but add complexity. Importantly, Scala does not address the problems with the Concept pattern (§2). In particular, it lacks model-dependent types and also precludes the dynamic dispatch that contributes significantly to the success of object-oriented programming [Aldrich 2013].

JavaGI [Wehr and Thiemann 2011] generalizes Java interfaces by reusing them as type classes. Like a type class instance, a JavaGI implementation is globally scoped, must uniquely witness its interface, and may only contain methods for the type(s) it is declared with. Unlike in Haskell, a call to an interface method is dynamically dispatched across all implementations. Although dispatch is not based entirely on the receiver type, within an implementation all occurrences of an implementing type for T must coincide, preventing multiply dispatching intersect across the Shape class hierarchy (cf. §5.1).

Approaches to generic programming in recent languages including Rust [Rust] and Swift [Swift] are also influenced by Haskell type classes, but do not escape their limitations.

Type classes call for a mechanism for implicitly and recursively resolving evidence of constraint satisfaction. The implicit calculus [Oliveira et al. 2012] formalizes this idea and extends it to work for all types. However, the calculus does not have subtyping. Factoring subtyping into resolution is not trivial, as evidenced by the reported stack overflow of the JavaGI compiler [Greenman et al. 2014].

No prior work brings type constraints to use sites. The use of type constraints as types [Wehr and Thiemann 2011; Swift] is realized as existentials in Genus. “Material–Shape Separation” [Greenman et al. 2014] prohibits types such as List<Comparable>, which do find some usage in practice. Existentials in Genus help express such types in a type-safe way.

Associated types [Myers 1995; Chakravarty et al. 2005] are type definitions required by type constraints. Encoding functionally dependent type parameters as associated types helps make certain type class headers less verbose [Garcia et al. 2003]. Genus does not support associated types because they do not arise naturally as in other languages with traits [Oliveira et al. 2010; Rust] or module systems [Dreyer et al. 2007] and because Genus code does not tend to need as many type parameters as in generic C++ code.

Finally, in Table 2, we compare how Genus and other languages perform with respect to the desiderata identified by prior work [Garcia et al. 2007; Siek and Lumsdaine 2011; Oliveira et al. 2010] and us. Not all prior desiderata are reflected in the table. Since we consider support for associated types to be an orthogonal issue, our desiderata do not include associated type access, constraints on associated types, and equality constraints (of which the latter are primarily used to express constraints on associated types). Also due to orthogonality, we omit type aliases and first-class functions.

11 Conclusion

The Genus design is a novel and harmonious combination of language ideas that achieves a high degree of expressive power for generic programming while handling common usage patterns simply. Our experiments with using Genus to reimplement real software suggests that it offers an effective way to integrate generics into object-oriented languages, decreasing annotation burden while increasing type safety. Our benchmarks suggest the mechanism can be implemented with good performance. Future work includes proving type safety of Genus and exploring more efficient implementations.
Table 2: Comparing various generics approaches

<table>
<thead>
<tr>
<th></th>
<th>C++11</th>
<th>SML/OCaml</th>
<th>Haskell</th>
<th>Java</th>
<th>C#</th>
<th>Cecil</th>
<th>C++ Concepts</th>
<th>Rust</th>
<th>Swift</th>
<th>Scala</th>
<th>⚘</th>
<th>JavaGI</th>
<th>Genus</th>
</tr>
</thead>
<tbody>
<tr>
<td>Multiparameter constraints (§3.1)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Multiple constraints (§3.2)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Associated types access (§10)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Retroactive modeling (§4)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Modular compilation (§5.1)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Implicit argument deduction (§6.2)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Modular type checking (§5.1)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Lexically scoped models (§4)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Concept-based overloading (§10)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Model generics (§3.2)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Natural models (§3.3, §4.4)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Non-unique modeling (§4.3)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Model-dependent types (§3.3, §4.5)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Run-time type/model info (§4.6, §7)</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
<td>☒</td>
</tr>
<tr>
<td>Model enrichment (§5.1)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Multiple dispatch (§5.1)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Constraint variance (§5.2)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Model inheritance (§5.3)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
<tr>
<td>Use-site generics (§6)</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
<td>☐</td>
</tr>
</tbody>
</table>

*aUsing MultiParamTypeClasses*

Acknowledgments

We thank Owen Arden and Chinawat Isradisaikul for their help with implementation problems, Ross Tate, Doug Lea, and Sophia Drossopolou for their suggestions about the design, and Tom Magrino for suggesting the name Genus.

This work was supported by grant N00014-13-1-0089 from the Office of Naval Research, by MURI grant FA9550-12-1-0400, by a grant from the National Science Foundation (CCF-0964409), by the German Federal Ministry of Education and Research (BMBF), grant 01IC12S01V, by the European Research Council, grant 321217, and by Quanta. The views and conclusions contained herein are those of the authors and should not be interpreted as necessarily representing the official policies or endorsement, either expressed or implied, of any sponsor.

References


A. Igarashi, B. Pierce, and P. Wadler. Featherweight Java: A minimal core calculus for Java and GJ. *ACM Transactions on Programming Languages and Systems*, 23(3):396–450, 2001.


