

A Survey on Abstract Data Types

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February 9, 2025

1 Introduction

From the categorization of species to the periodic table of elements, the organization and manipulation of data is integral to all aspects of science and technology. Although the collection of data is not a new field of research by any means, the advent of computerized systems, with its ability to transfer and use massive amounts of data in mere seconds, necessitates understandable and sensible organization structures.

Over the past 60 years, as computational resource restrictions have eased dramatically, programming languages and software creation tools have evolved with increasingly better methods of dealing with data. By viewing pieces of data through specific lenses, programmers can then write algorithms that operate on any data matching the shape of those lenses, rather than the data itself. These lenses are more commonly referred to as data types, which assign a type to a piece of data. As we will see later in this paper, further abstractions over data types themselves enable extremely concise and readable code, both for the programmer and computer.

The rest of this article will explore data type concepts from `¡X¿` languages, structures to handle errors that might appear, and a small review of ergonomics and usability of those data types. A necessary discussion of various type systems will naturally evolve as a precursor to data types themselves.

2 Data Types in `¡X¿` Languages

Abstract Data Types (ADTs) can look quite different in many languages, but often share the same principles and rules that make it understandable for programmers. In its essential form, an ADT is a collection of data together with operations that can be carried out on that data [21]. The key differences between ADTs from different languages are the conventions used to collect data and the power operations have over collected data.

In popular object-oriented programming (OOP) languages, ADTs are approached with a “class” mindset, where programmers specify a blueprint for objects that will eventually be used in a program. For better code reuse, classes can be built from pre-existing classes through the use of inheritance, a term used

to express that a new class inherits all the properties of the old class. Objects created by classes are referred to as instances of the class and often hold some amount of mutable data, with methods that have full access to the data. The methods can choose to make use of “side effects”, referring to the modification of data not explicitly passed through regular function arguments or return values. Side effects in methods have the consequence that the output of a method may change independent of the values passed through its arguments. In other words, the object a method has access to may contain some state that influences the outcome of the method.

At the opposite end of the programming world lies purely functional languages. ADTs in these languages are more often referred to as data types rather than classes and are only used for collecting and operating on data. “Objects” created from these data types are immutable and methods are instead replaced with pure functions that explicitly take in the relevant data type as input. The operators attached to a functionally pure ADT are functionally pure themselves, meaning they cannot make use of any hidden state and therefore cannot have side effects.

2.1 Python

As a dynamic interpreted scripting language with thousands of high quality libraries and a short learning curve, Python has become one of the most dominant modern programming languages across several subdomains of computing. The language itself is characterized mainly by its simple approach to writing readable, practical, and user-friendly code [13], used by many as a “glue” to easily join high performance computing libraries. The role of Python as a mediator between optimized C libraries and human understanding, especially in the machine learning space, is adjacent to the role of documentation in expressing the intention and meaning behind code, rather than focusing on the small details of the code itself.

2.1.1 Python and Types

The overarching data model behind Python is that everything is either an object or a relation between objects. Almost all objects in Python are used by reference, and each object contains an identifier (memory address), an immutable type, and then the actual value itself. In contrast to other scripting languages (namely JavaScript), an immutable type means that Python will never implicitly recast the type of an object to fit the expected types of functions or methods (ie. it is strongly typed).

Although Python is strongly typed, its lack of a static typing system serves as a double edged sword for readability and reliability. With small programs, some developers may prefer to just “get things done”, making use of whatever combination of builtin types are available. A mixture of sets, dictionaries, and lists with unrestricted element types are all too common in these types of scripts. However, assigning proper data types to work with various pieces of information

is essential to the longevity and maintenance of large scale projects. Since the introduction of optional static type hints via a library in Python 3.5 and increased support through Python 3.8, a 2022 survey [4] found a steady increase in the number of GitHub repositories making use of static type hints.

Aside from granting linters more information to statically analyze the program and warn developers of possible errors, types also provide programmers themselves with useful information regarding their program. As reinforced by the philosophy of TypeScript as a type-hinted version of JavaScript, data types serve as an important channel of communication between programmers. An additional finding of the 2022 survey [4] showed that typecheck errors in the linters did not prevent a large amount of code from being pushed, but were still found to be useful outside of the linter.

Fundamentally, Python places little restriction on the programmer's actions. For example, while data privacy within Python objects is theoretically possible through name mangled fields (beginning with double-underscores), there is no real way to completely restrict access to an object's fields so long as that object exists in the code. Access to deeper implementation details, like a list of current variables, list of fields in any variable, and even interpreted bytecode is possible through the use of magic methods and variables (denoted by two underscores at the beginning and end of the identifier).

2.1.2 Example: A Linked List

Since Python is primarily an OOP language, the most common form of expressing data types is through the use of classes. A typical linked list data type in Python 3.11 would look something like:

```
from __future__ import annotations
from typing import Generic, TypeVar

T = TypeVar("T")

class LinkedList(Generic[T]):
    def __init__(self, value: T, next: LinkedList[T] | None = None) -> None:
        self.value = value
        self.next = next

    def append(self, new_value: T) -> None:
        l = self
        while l.next is not None:
            l = l.next
        l.next = LinkedList(new_value)

    def concat(self, other: LinkedList[T]) -> None:
        l = self
        while l.next is not None:
            l = l.next
        l.next = other
```

Note that Python 3.12 and later make some major changes involving scoping and the syntax of generic types [14].

The above `LinkedList` class expresses the recursive linking nature of the data type while demonstrating commonly used side effects of OOP methods. Note that the return type for both `append` and `concat` is `None`. Instead, the data contained within the `self` object is mutated in place as needed to achieve the desired functionality.

2.1.3 Pydantic: A data validation library

While native Python works well for simple data types like linked lists, type hints do not provide any guarantees that the type of value, for example, will be consistent across all elements in the list. So, while it does serve as a useful notation tool for documenting the structure of a data type to other programmers, there are no insurance that the runtime values for these objects are consistent with the type annotations. As with many topics in Python, a convenient library by the name of Pydantic will enable both type hints and data validation for those types *at runtime*. The Pydantic version of our linked list will now look something like:

```
import pydantic

from __future__ import annotations
from typing import Generic, TypeVar

T = TypeVar("T")

class LinkedList(pydantic.BaseModel, Generic[T]):
    value: T
    next: LinkedList[T] | None = None

    def append(self, new_value: T) -> None:
        l = self
        while l.next is not None:
            l = l.next
        l.next = LinkedList(value=new_value)

    def concat(self, other: LinkedList[T]) -> None:
        l = self
        while l.next is not None:
            l = l.next
        l.next = other
```

For this simple data type, only the init function changes. However, under the hood, Pydantic will now ensure that the type of `value` within this linked list remains consistent across all elements, throwing an exception otherwise. Pydantic's version of a class has the ability to transform an OOP construct into one more closely related to functional programming. If desired, the use of decorators above field names or methods (denoted by `@decorator`) can restrict mutability, modify constructors, and even change the data validation process to ensure objects are of the correct type.

2.1.4 Conclusion

Python's high-level nature, excellent documentation, library support, and exposure of underlying architectures when needed combine to create a powerful, flexible, and productive language. Although many aspects of the language are initially hidden or abstracted for the sake of convenience, Python allows programmers to dig deeper into the implementation when necessary and extend the language in a manner best suited for a specific task.

2.2 TypeScript

In JavaScript, types are very loosely bound, and many operations/functions will silently convert types at runtime. The silent recasting of JavaScript objects combined with unintuitive defaults can result in very confusing results for the uninitiated. For example, the expression `1 + "1"` returns a value of 11 and the expression `[5, 11, 2].sort()` returns `[11, 2, 5]`. In both expressions, the numerical values are converted into strings before the operation is performed.

TypeScript, as an extension of JavaScript with types, is similar to Python in the sense that both languages started with dynamic typing and patched on static types after the fact, and both aim to resolve issues caused by dynamic typing. TypeScript (and Python) use types not only for more assurance in program correctness, but also for the sake of convenience. When making full use of the available typing systems, text editors and IDEs are able to assist in new code development through intellisense and autocompletion of methods and variables. Types of objects and documentation for those types additionally appear on mouse-hover, providing fast in-context knowledge.

The below list outlines key properties of types in TypeScript.

- Types are treated as sets of values, where types can be part of multiple sets, a union of sets, or rarely even an intersection of sets.
- Types are structural. Any object containing the required fields needed to be a member of a certain type will be considered an object of that type, even if a different type name has been assigned to it. Further, if an object has additional fields not specified by the type, the extra fields will be ignored in the typechecking process. Essentially, as long as an object meets the minimum fields required by the specified type, the object is assumed to be a member of that type.
- TypeScript allows any literal string to act as a subset of its respective type, with the only allowed value being the literal itself. For example, a field with the type `'name'` will only accept the value of `"name"`. However, a field with the type `String` will accept any string object.

In comparison to algebraic data types often seen in functional languages, TypeScript replaces inductive data types with discriminated union sets. For example, our linked list type may look something like:

```

type LinkedList<T> =
  | { kind: "nil" }
  | { kind: "cons"; value: T; next: LinkedList<T> };

function append<T>(self: LinkedList<T>, new_value: T): LinkedList<T> {
  if (self.kind === "nil") {
    return { kind: "cons", value: new_value, next: { kind: "nil" } };
  }

  let l = self;
  while (l.kind === "cons" && l.next.kind !== "nil") {
    l = l.next;
  }
  l.next = { kind: "cons", value: new_value, next: { kind: "nil" } };
  return self;
}

function concat<T>(self: LinkedList<T>, other: LinkedList<T>): LinkedList<T> {
  if (self.kind === "nil") {
    return other;
  }

  let l = self;
  while (l.kind === "cons" && l.next.kind !== "nil") {
    l = l.next;
  }
  l.next = other;
  return self;
}

```

Because TypeScript allows literals to act as types, the only valid values in the “kind” field of an object represented by the `LinkedList` type are our constructor names: “nil” or “cons”. Although TypeScript treats every object as a reference, similar to Python, we have not used side effects in the `append` or `concat` functions. Instead, we explicitly accept the linked list we want to transform as input, perform the required action, and return the head of the linked list in the output.

For a non-inductive approach to the `LinkedList` data type, we can use TypeScript’s interfaces to create something that looks much more like the Pydantic version:

```

interface LinkedList2<T> {
  value: T;
  next: LinkedList2<T> | null;
}

function append2<T>(self: LinkedList2<T>, new_value: T): LinkedList2<T> {
  let l = self;
  while (l.next !== null) {
    l = l.next;
  }
  // Structural typing - No need to refer to LinkedList2 name explicitly
  l.next = { value: new_value, next: null };
  return self;
}

```

```

function concat2<T>(  
    self: LinkedList2<T>,  
    other: LinkedList2<T>,  
) : LinkedList2<T> {  
    let l = self;  
    while (l.next != null) {  
        l = l.next;  
    }  
    l.next = other;  
    return self;  
}

```

Finally, a pure OOP version of our `LinkedList` type, using classes, can be formed. Note that the `append` and `concat` functions are now methods with side effects:

```

class LinkedList3<T> {  
    value: T;  
    next: LinkedList3<T> | null = null;  
  
    append3(new_value: T) {  
        let l: LinkedList3<T> = this;  
        while (l.next != null) {  
            l = l.next;  
        }  
        l.next = new LinkedList3();  
        l.next.value = new_value;  
    }  
  
    concat3(other: LinkedList3<T>) {  
        let l: LinkedList3<T> = this;  
        while (l.next != null) {  
            l = l.next;  
        }  
        l.next = other;  
    }  
}

```

2.3 Rust

2.4 Dafny

2.5 Haskell

2.6 Agda

2.7 APL

APL is an extremely unique language that prides itself on ultra-concise expressions and data-based SIMD parallelism. Among the most defining characteristics of APL is the tendency to make use of non-ascii symbols for operations, right-to-left operator associativity, and the fact that everything in APL is array-based. APL statements are often branchless in nature, and mapping functions over data is strongly preferred over regular loops.

One APL reference [9] compares the types within APL to the numpy library in Python. Thus, instead of providing syntax to create data types as seen in other languages, APL strongly encourages the use of arrays and matrices to solve problems. The shape of the array (rank and depth) determines the type of an object and the effects of applied operators.

Because APL only has one major data type (the array), operators are often composed without explicitly calculating intermediary values. APL categorizes all operators into two groups: monadic operators that take in one argument, and dyadic operators for two arguments. More explicitly, all operators must accept either one or two arguments, which may drastically change the operator's behaviour. As a simple example, the monadic tilde operator ($\tilde{}$) is used for boolean *not*, while the dyadic version is for set difference.

Given the knowledge that everything is an array, we cannot reasonably recreate our linked list example. Nor should we, since lists and vectors are first-class citizens in APL. Instead, we take a look at a more interesting aspect of APL's feature set - the avoidance of branching through bit masks. Consider the following "loop and branch" Python code, where we loop over an array of objects and return every object that satisfies some function:

```
def select(f, array):
    """Loop & branch"""
    result = []
    for element in array:
        if f(element):
            result.append(element)
    return result
```

The concise APL equivalent would instead be the following, notably branchless, code:

```
elems ← f array
elems/array
```

Note that the first line maps a function $f : T \rightarrow \text{bool}$ to create a bit mask (an array of zeroes and ones) *elems*, which is then passed to the selection operator $/$. Since both implementations must apply f to all elements of the input array, the branchless version can be far more efficient due to the use of SIMD instructions.

2.8 Unified Modelling Language (UML)

Up until this point, all the languages explored in this document have been targeted at making computers execute a desired set of instructions over some type of data. Of course, the specific methodology and concepts used to achieve this human-computer interaction differ greatly between languages, with each maintaining their own idioms and customs. Conversely, UML is not a programming language in the normal sense, but a standardized method of visualizing and clearly communicating the intentions behind a system.

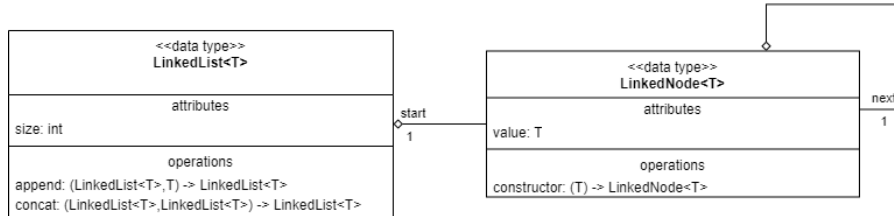
According to the UML documentation, UML is intentionally independent of any and all processes that could be used to actually implement a system.

UML diagrams are not intended to comprehensively cover every aspect of a system's implementation, including different quirks and features of different languages, but rather should be used as a powerful aid in the creation of program documentation. Generally, different types of diagrams are used for recording types, state, information flow, and even the execution of the program itself.

In the context of UML, a data type is a specialized form of classifier with no internal structure. Each data type may contain several named fields to contain data, along with operations on that data. The focus of data type based diagrams is to convey the following to programmers:

1. All data types available to the user.
2. All relationships between data types, including inheritance, extensions, use of types within the fields of other types, and even type usage in operators.

Below is an example of our famous LinkedList type, modelled by UML:



A simple standalone type like LinkedList may not be complex enough to showcase the full advantages of UML, but it is enough to demonstrate the effectiveness of visualization in understanding code. From the arrows in the image, it is immediately obvious that the start attribute of LinkedList requires a ListNode and ListNodes may build off of themselves indefinitely.

UML diagrams are a simple and concise method for conveying the meaning and intention behind a system. While the syntax is largely standardized, because this language is meant for communication with humans rather than machines, slight modifications of syntax for the sake of clarity can be made as necessary. For example, the types of operators may be omitted for an ultra-compact diagram, or kept to ensure an agreement on the interface implementation is met.

2.9 Event-B

Although UML is a powerful tool for communicating the relationship between data types, the visual methodology falls short when attempting to convey the actionable properties of a system. In continuing the exploration of modelling languages, we find that the Event-B model not only encompasses actionable system traits, but does so with consideration of the system's environment. Therefore, Event-B is suitable for formal system analysis and proving the correctness of software following the model specification.

Data types for Event-B may not exist in the normal sense of the word, but a different approach to data abstraction through the concepts of machines, contexts, and events is more than suitable for the same purpose:

- Machines are somewhat adjacent to classes in OOP, acting like miniature programs with some provable traits consisting of variables, invariants, theorems, a single variant, and a list of events that may cause the machine to “run” (analogous to an object’s methods). Like class inheritance, new machines can “refine” old ones by carrying over and adding new entries to the aforementioned traits.
- Contexts, on the other hand, are much closer the formal definition of abstract data types, defining a set of axioms, theorems, constants and data in the form of carrier sets. Contexts can be extended in a similar fashion to class inheritance, with the child context carrying over all properties of the parent context.
- Finally, events serve as activation conditions for the machine they are assigned to. They introduce possibly nondeterministic actions into the system in a formal manner, and must clearly state how they affect a machine’s variant. A key aspect of events is that all actions within an event are performed simultaneously. This means that, unlike in sequential programming languages, the order in which actions are defined have no relevance on the outcome of the event. Of course, the restriction of simultaneous execution means that users of the language cannot reassign variables in more than one action, nor should they rely on the deterministic nature of normal program execution.

All expressions in the above abstractions use formal discrete math notation.

Although a modelling language focused on provable aspects of a system is not very suitable for a simple low-level data type, we nonetheless attempt a simplified implementation of `LinkedList` with only the `append` method.

<pre> LinkedListContext sets Nodes // Set of nodes in the list References // Set of references constants null // End of list axioms axm1: References = Nodes \cup {null} axm2: $\forall n : n \in Nodes \implies n.next \in References$ // All nodes must have a reference </pre>

```

AppendEvent
  any
    last_node
    new_node
  where
    last_node  $\in$  Nodes
     $\{last\_node\} = \{node \in Nodes : node.next = null\}$  // last_node is the
end of the list
    new_node.value : T // New value must have the same type as existing
nodes
    new_node.next = null // New node is end of the list
  then
    last_node.next := new_node // Last node is no longer the last node
    Nodes := Nodes  $\cup$  new_node // Add node to list

```

```

LinkedListMachine
  variables
    head
    value
    next
  invariant
    inv1: head  $\in$  Nodes
    inv2: value : T // Generic type T
    inv3: next  $\in$  References
    inv4:  $\forall n : n.next \neq null \implies n.next.next \in References$  // Nodes
can either end or continue the list
  events
    AppendEvent

```

The latest Event-B wiki [5] notes future plans to support recursive structured types like inductive LinkedLists in a more concise format.

2.10 Exploration of Conceptual Data Types

2.10.1 The Axiomatic Approach to Data Types

As stated in the introductory textbooks [11] and [21], ADTs at their core are just collections of data coupled with operations. Both textbooks introduce data types in a manner not unlike the ideas of Event-B, using formal notation to guarantee precise properties of the given data type. Interestingly, both textbooks opt to separate data type operations into syntactic and semantic components, where the semantics contain a list of axioms to be fulfilled by each operation. Note that these axioms are almost entirely removed from the implementation of the type, acting more as a specification of requirements rather than a how-to guide.

Guaranteed type completeness and correctness are two key features of the axiomatic approach. That is, axioms encourage operations to consider all pos-

sible values that the type may take, especially regarding edge cases and errors. For example, in our linked list, we need to ensure operations over an empty list are well-defined and that we can create the empty list itself in the first place. Of course, in the case of linked lists, this is near-trivial, but focusing on the edge cases in this manner can guarantee correctness of far more complex types.

Consider the axiomatic definition of a linked list, according to the specification of [21]:

```

NAME
    LinkedList(item)
SETS
    N: Set of nodes
    I: Set of items
SYNTAX
    new :→ N
    append : N × I → N
    concat : N × N → N
    deleteLast : N → N
    tail : N → I
    isEmpty : N → bool
SEMANTICS
    ∀i ∈ I, ∀n, n1 ∈ N :
        deleteLast(append(n, i)) = n
        tail(append(n, i)) = i
        tail(concat(n, n1)) = tail(n1)
        isEmpty(new()) = True
        isEmpty(append(n, i)) = False
        tail(new()) = error // New initializes a list but does not hold any
value
        deleteLast(new()) = error // Cannot delete from an empty list

```

The above axioms do not express the behaviour of each function independently, but instead focus on relationship between one another. For instance, the axiom $\text{tail}(\text{new}()) = \text{error}$ outlines two properties with one axiom:

1. Finding **tail** on an empty list produces an error.
2. Since **tail** produces an empty list error, we must know that **new** creates an empty list.

Similarly, the axiom $\text{deleteLast}(\text{append}(n, i)) = n$ shows that **append** must add the item i to the end of the list n , in order for **deleteLast** to perform correctly.

Unfortunately, this approach may present drawbacks for implementing unknown data types in popular programming languages *because* of its coupled operation behaviour. In many programming languages, especially non-functional ones, developers need to implement each data type operation mostly independent of other operations. Therefore, ensuring congruency between written code and the given specification presents a new challenge for programmers; data types

becoming lost in the translation from dependent specification to independent implementation serve as an undesirable source of bugs and mistakes. A compiler for such a language would greatly reduce friction between theory and application, but the semantic expressivity available from the axiomatic approach may be worth the costs.

2.10.2 A More Constructive Approach

The textbook [21] specifies an additional method of data type specification focused on construction of new objects from old ones. Instead of axioms, each operation can specify relevant pre-conditions and post-conditions. This method is far closer to the implementation of data types in programming languages, with an uncanny resemblance to the optional input restrictions seen in Dafny functions. For conciseness, the example given in Section 2.4 serves as a good approximate illustration of the constructive approach outlined in the textbook.

3 Error Handling and Undefinedness

4 A Word on Usability

5 Conclusion

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