Automated Verification of Data Properties and Linearizability for Heap-Manipulating Programs

CONG QUY TRINH
Abstract


Software verification is the process of verifying a software application by checking whether it satisfies requirements. The automated verification of programs is one of the most challenging problems in software verification. The problem becomes even more challenging when dealing with concurrent programs that dynamically allocate memory on the heap. In this thesis, we propose methods for verifying both safety, shape, and linearizability properties of both sequential and concurrent heap-manipulating programs. In short, linearizability means that concurrent operations appear to be executed atomically on a single machine. Such programs induce an infinite-state space in several dimensions: they consist of an unbounded number of concurrent threads, an unbounded number of pointers, they use unbounded dynamically allocated memory, and the domain of data values is unbounded. In addition, we verify linearizability properties of concurrent programs whose linearization points are either fixed or depend on the future execution of the program. In this thesis, we describe three approaches for verifying safety, shape, and linearizability properties.

In the first approach, we present a framework for verifying programs that manipulate dynamic linked data structures, whose correctness depends on ordering relations between data values. Our framework extends that of forest automata, in which the heap is described by a set of tree automata, by adding data constraints that express relationships between data elements associated with cells of the heap. This approach works for verifying safety property of sequential programs.

In the second approach, we present a framework for automatically verifying linearizability of concurrent data structures with an unbounded number of threads. In this framework, non-fixed linearization points (LPs) are handled by asking the user to specify so-called linearization policies, which are mechanisms for assigning LPs to executions. To handle an unbounded number of threads, we use the thread-modular approach which allows to bound the number of considered threads. To handle an unbounded heap, we define an abstraction, which precisely describes the parts of the heap that are visible (reachable) from global variables, and makes a succinct representation of the parts that are local to threads. We have applied the framework to prove linearizability for a wide range of concurrent data structures based on singly-linked lists.

In the third approach, we present a novel shape analysis that can handle heap structures which are more complex than just singly-linked lists, in particular we handle skip lists and arrays of singly linked lists, while at the same time handling an unbounded number of concurrent threads, an unbounded domain of data values (including timestamps), and an unbounded shared heap. Our approach represents a set of heaps by a set of so-called fragments. A fragment is an abstraction of a pair of heap cells that are connected by a pointer field. To the best of our knowledge, our framework is the first that can automatically verify concurrent data structure implementations that employ singly linked lists, skiplists as well as arrays of singly linked lists, at the same time as handling an unbounded number of concurrent threads, an unbounded domain of data values (including timestamps), and an unbounded shared heap.

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Dedicated to my parents and my wife and daughter
List of papers

This thesis is based on the following papers, which are referred to in the text by their Roman numerals.

I  **Verification of Heap Manipulating Programs with Ordered Data by Extended Forest Automata**
Parosh Aziz Abdulla, Lukáš Holík, Bengt Jonsson, Ondřej Lengál, Cong Quy Trinh, Tomáš Vojnar.

II  **Automated Verification of Linearization Policies**
Parosh Aziz Abdulla, Bengt Jonsson, and Cong Quy Trinh.
In this thesis, the paper has been extended with appendices that could not be accommodated in the published version.

III  **Fragment Abstraction for Concurrent Shape Analysis**
Parosh Aziz Abdulla, Bengt Jonsson, and Cong Quy Trinh.
In this thesis, the paper has been extended with appendices that could not be accommodated in the published version.

I am the main author of papers I, II, and III. The ideas originated and were developed in discussions with other authors. I am the sole implementor of I, II, and III. I wrote the papers together with other authors. Reprints were made with permission from the publishers.
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1. Introduction

Computers have been used for a variety of applications in business, science, education, engineering and so on. They help to solve real-world problems that would otherwise be slow, impossible or extremely difficult to address without computers and software. However, sometimes they do not behave exactly as we expect them. In many cases, the consequences could be very serious, for example when errors in banking or flight control software result in unexpected behaviours. Errors in computer systems are mostly not caused by the machine itself, but typically originate from the software that controls the computer systems, so-called bugs. Bugs are quite common in complex software systems since such systems typically have complicated input and involve many features, making them difficult to design and implement by human effort. Detecting and fixing software bugs are important tasks in software development. They can be very hard to detect and correct, especially if they are discovered after the software has been delivered. Therefore, it is very important to allocate sufficient resources, both in terms of time and manpower, to ensure that developed software is as free of bugs as possible.

Some bugs are less serious than others. Some types of software, e.g., in user interfaces or entertainment software, can be usable even if it contains a small number of bugs. However, in the case of critical systems and system components such as software in libraries of programming languages, bugs can have far-reaching consequences, and must be avoided as much as possible. Some libraries provide implementations of standard data structures such as stacks, queues, and other containers. Such data structures provide ways of storing and retrieving data in a way that suits the application at hand. A data structure should ideally provide a simple interface to the software that uses it. An interface provides the set of operations with specifications about their types of arguments and returned values. For example, a stack allows inserting and removing elements in a particular order. Every time an element is inserted, that element is removed in reverse order of insertion. The simplest application of a stack is to reverse a word. You insert a given word to stack - letter by letter - and then remove letters from the stack. By using library data structures, data can be easily and efficiently exchanged; it allows portability, comprehensibility, and adaptability of information.

A data structure can be sequential or concurrent, where concurrent is tricky and difficult to get correct. Concurrent data structures that can be accessed and manipulated concurrently by many parallel threads are a central component of many parallel software applications. Data structures typically use
heap-allocated memory to store their data. For example, the concurrent linked queue in java.util.concurrent uses a singly linked list to organise data. The data structure implementation may be quite complex using skiplists and binary trees.

1.1 Formal Verification

The predominant method to ensure software quality is testing. It is a dynamic analysis where a program is executed under specific conditions, in so-called test cases, while checking whether the result for a given input matches the expected output. The test cases should be carefully designed to cover as many as possible cases of program executions. However, it is infeasible to cover all possible executions. Therefore, it is said by Edsger W. Dijkstra that testing can be used to show the presence of bugs, but never to show their absence. It would be nice to have techniques for checking that all executions conform to the interface of a data structure. A possibility is to use formal techniques which is the approach used in this thesis.

Formal verification uses mathematical methods to check whether a program, or piece of software, satisfies its specification. There are several approaches to formal verification, including equivalence checking [?], theorem proving [107], and model checking [40]. Equivalence checking decides whether a system is equivalent to its specification with respect to some notion of equivalence. In industry, this is mostly used for hardware designs. Theorem proving is a technique where both the behavior of the system and its desired properties are expressed in mathematical logic. Then, theorem proving, typically assisted by an interactive theorem prover, will try to prove that the system satisfies these properties.

Model checking takes as input a model of the program under consideration and a formal specification of a property to be verified. The specification of a software component may consist of a number of such properties, each of which can be verified using model checking. The approach exhaustively explores all possible executions of the model. This is typically done exploring the set of reachable states of the model which can be finite or infinite. This works well if the set of reachable states is finite which typically happens for embedded controllers and hardware designs. However, most softwares are infinite-state, e.g., a data structure may contain an unbounded amount of data. A common technique for handling this is to devise a symbolic representation of sets of states, such that a single symbolic representation represents an infinite set of states. However, it is difficult to find a suitable symbolic representation for data structures, in particular, complicated data structures such as trees and skip-lists, where the relationships between heap cells are complicated in both reachability and data aspects.
1.2 Research Challenges

The challenge addressed in this thesis is to develop techniques for automated verification of both sequential and concurrent data structures that employ dynamically heap-allocated memory. This requires to address several challenges in automated verification:

- **Dynamically heap-allocated memory**: Data structures typically use dynamically heap allocated memory. In each cell of a heap, the domain of data values can be unbounded. In the area of formal verification, there exist several approaches for heaps and for data, but there are few approaches for combining them in suitable ways. In this thesis, we provide an approach to automated verification of sequential data structures where correctness depends on relationships between data values that are stored in the dynamically allocated structures. Such relations on data are central for the operation of many data structures such as search trees, priority queues (based, e.g., on skip lists), key-value stores, or for the correctness of programs that perform sorting and searching. There exist many automated verification techniques dealing with these data structures, but only few of them can automatically reason about data properties. They are often limited to specific classes of structures, mostly singly-linked lists (SLLs). Our approach is based on the notion of forest automata [64] which has previously been developed for representing sets of reachable configurations of programs with complex dynamic linked data structures.

- **Unbounded number of threads**: For the case of concurrent data structures, we have to verify that the data structures are correct for any number of threads that access and manipulate the structures. We handle this challenge by extending the successful thread-modular approach[58] which verifies a concurrent program by generating an invariant that correlates the global state with the local state of an arbitrary thread. In this approach, we only verify each thread separately using an automatically inferred environment assumption that abstracts the possible steps of other threads.

- **Specification of correctness**: To ensure that a concurrent data structure is correct, we have to specify a correctness criterion that relates the concurrent interface to the interface of a corresponding sequential data structure. One such correctness criterion is linearizability[75]. Linearizability is generally accepted as the standard correctness criterion for concurrent data structure implementations. Intuitively, it states that each operation on the concurrent data structure can be viewed as being performed atomically at some point (called linearization point (LP)) between its invocation and return. Existing approaches for verifying linearizability of concurrent data structure implementation are limited to specific classes of concurrent data structures based on simple heap
structures such as singly linked lists; so far no technique (manual or automatic) for proving linearizability has been proposed that is both sound and generally applicable. In this thesis we provide a technique to specify linearizability of concurrent data structures.

- **Unbounded number of pointers**: Some data structure implementations can dynamically allocate memory cells with an unbounded number of pointer fields, such as cells in skip-lists and arrays of lists. It is difficult to provide a symbolic representation for such data structures. There are no techniques that have been applied for automatically verifying concurrent algorithms that operate on such data structures. We address this problem by proposing a technique called *fragment abstraction* in which a heap is divided into small pieces called fragments. A fragment is an abstraction of a pair of heap cells that are connected by a pointer field. Our approach is general and precise enough to verify the above complicated data structures.

The following sections are organized as follow: in Section 2, we present the general background about model checking, then in Section 3, we describe data structures. Thereafter, in Section 4, we describe our approach to specify linearizability. Our heap abstraction techniques are described in Section 5. Finally, in Section 6, we summarize and give future plans for our work.
2. Model Checking

The approach that we focus on this thesis is model checking. Automation

This approach was introduced by Emerson and Clarke [41] and by Queille and Sifakis [105]. Model checking aims to check whether a model of a program satisfies a given specification. The method then computes and returns either "correct" when the specification is satisfied by the program, or "incorrect" when the program does not satisfy its specification. In the latter case, the method can explain the reason by giving a counter-example. Models are typically transition systems consisting of states and transitions between states. A state in the model contains relevant information about the program. Alongside all the states of the system, the model also depicts the transitions, i.e. how to move from one state to another state. Every behaviour of the system is represented as a succession of transitions, starting from some initial state. The number of states and transitions can be finite or infinite. A specification consists of properties of behaviors (i.e., of sequences of states). One usually distinguishes safety properties ("something bad must never happen") from liveness properties ("something good must eventually happen"), where most properties are safety properties. Model-checking aims to explore the state-space entirely from some initial states. One of the main problems with model checking is that the state-space is infinite or finite and very large. It grows in-fact exponentially with the number of parameters or the size of their domains. Therefore, there have been several methods to address the state-space explosion problem.

There are several techniques addressing the state-space explosion problem. One important approach is partial order techniques, which aim to avoid redundant exceptions. This approach is based on the fact that, in some cases, exploring all orderings of transitions that are independent is not necessary. Perhaps the main approach to address the state-space explosion problem is to use a symbolic representation. It avoids representing concretely all states of the system. Its main idea is to design a symbolic representation of sets of states. This is done by dropping irrelevant details based on properties that we want to verify. Symbolic representations are of crucial help to combat the state-space explosion, accelerate the algorithms, and get them to terminate in a reasonable amount of time.
Example:

Dining Mathematicians

\[ n : \text{integer} \]

assume \( n \geq 1 \)

Let us illustrate symbolic representation by showing an example that consists of two threads representing mathematicians, which share an integer variable \( n \). The number of states is infinite. Suppose we want to check that the two mathematicians are not eating at the same time, i.e., we want to check the invariant that \( eat_1 \land eat_2 \) will never happen. Simple inspection shows that this invariant indeed holds for the system, for the simple reason that when the left process is in \( eat_1 \), then \( n \) is odd, and when the right process is in \( eat_2 \), then \( n \) is even. However, this simple fact cannot be detected by naive reachability analysis, since the system has infinitely many states. The dining mathematicians example is of course simple and does not reflect the complexity of today’s software.

Finding a right symbolic representation is challenging, it might introduce behaviours that could turn out to be bad. Then, the method would return that the property is not satisfied and we would not know whether it comes from the approximation introduced by the symbolic representation or from the concrete system itself. For example in the dining mathematicians problem, if we would start with an over-approximation that ignores the variable \( n \), we get a false positive, we must then refine the symbolic representation to avoid it. To deal with the imprecision caused by a too coarse over-approximation, it is possible to analyse the returned counter-example and find the origin of the problem. If it turns out to be a real concrete example, the method has in fact found a bug, and the property is surely not satisfied. Otherwise, the counter-example
comes from the approximation, that is, there is a step in the sequence of events leading to that counter-example which is not performed by the original system but only by the abstract model. The approximation must then be refined to discard this step and the method should be run anew.

Nevertheless, finding suitable over-approximations is a challenge on its own. This thesis revolves around the problems of unboundedness, including dynamically heap-allocated memory, unbounded number of threads, and unbounded number of pointers which are described as research challenges in the previous section.
3. Data Structures

In general, a data structure is any data representation and its associated operations. Even an integer or floating point number stored on the computer can be viewed as a simple data structure. Typically, a data structure is meant to be an organization or structuring for a collection of data items. Each data structure has an interface which defines a set of possible values and a set of operations. More precisely, the interface consists of a set of operations or methods, each having a number of input and output parameters, and a specification of the effect of each operation.

For example, a sequential set is an data structure for storing a collection of elements, with the following three operations:

- \text{add}(e) \text{ adds element } e \text{ into the set, returning true if and only if } e \text{ was not already there.}
- \text{remove}(e) \text{ removes element } e \text{ from the set, returning true if and only if } e \text{ was there.}
- \text{contains}(e) \text{ checks the existence of element } e \text{ in the set, returns true if and only if the set contains } e.

For each method, we say that a call is successful if it returns true, and unsuccessful otherwise. It is typical that in applications using sets, there are significantly more \text{contains()} calls than \text{add()} or \text{remove()} calls. The implementation of a data structure should provide an efficient way to store data in computer memory and perform its operations in an efficient way. Data structures typically use heap-allocated memory to store their data. Various schemes can be used to organise the heap-allocated memory, such as singly-linked lists, doubly-linked lists, skip-lists and trees. The implementation can be sequential or concurrent.

A set can be implemented as a singly linked list of cells. Each cell has two fields named val and next. The val field represent the value of a cell, and the next field is a reference to the next cell in the list. Cells are sorted according to val order, providing an efficient way to detect when an item is absent. The list has two sentinel cells, called head and tail, which are the first and last list elements. Sentinel nodes are never added, removed, or searched for, and their values are the minimum and maximum integer values.

A concurrent data structure is a way of storing and organising data for access and manipulation by multiple computing threads (or processes) on a shared-memory computer. Each operation is implemented as a sequential method that is executed by a thread accessing the share state. Several features of shared-memory multiprocessors make concurrent data structures significantly more difficult to design and to verify correct than their sequential...
counterparts. The primary source of this additional difficulty is concurrency: because threads are executed concurrently, possibly on different processors, and are subject to operating system scheduling decisions, interrupts, etc., we must think of the interaction between threads as completely asynchronous, so that the steps of different threads can be interleaved arbitrarily.

There are several techniques to construct concurrent data structures including coarse-grained locking, fine-grained locking, and lock-free programming. The simplest technique is coarse-grained locking, where a single lock is used to synchronise every access to an object. Coarse-grained locking is easy to reason about; however it works well only when the level of concurrency is low. However, if too many threads try to access an object at the same time, then the object becomes a sequential bottleneck, forcing threads to wait in line for access. Fine-grained synchronisation techniques address this problem by splitting the object into independently synchronised components, ensuring that method calls interfere only when trying to access the same component at the same time. Fine-grained locking requires very careful design of the data structure and its methods, since one must foresee what can happen when several threads access the same component in parallel. Fine-grained synchronisation is often performed without locks, replacing them by less costly synchronisation operations such as `compareAndSet()`. Each of these techniques can be applied (with appropriate customisation) to a variety of common data structures (queues, stacks, sets) implemented by different linked data structures such as singly linked lists, skiplists, trees, or lists of lists.

As an example, Fig. 3.1 depicts a program Lazy Set [70] that implements a concurrent set containing integer elements with three operations `add`, `remove` and `contains`. It is just as the sequential version, except that each cell now has two additional fields named `mark` and `lock`. The field `mark` is true if the node has been logically removed from the set. The `lock` field is a lock and the field `val` represents the data value which in this case is an integer. The mechanism behind logically and physical removing is explained as following: it is impossible to atomically remove a cell from the list if other threads may concurrently access the adjacent cells. One reason is that one must both move a `next` pointer which references to the cell as well as physically remove the cell. This cannot be done, e.g., if another thread currently is visiting the cell that is to be removed. Therefore, the task of removing a cell from the list can be split into two phases: the cell is logically removed simply by setting a `mark` field to be true, and later, the cell can be physically deleted by unlinking it from the rest of the data structure. The removal “actually happens” when an entry is marked, and the physical removal is just a way to clean up. The algorithm uses two global pointers, `head` that points to the first cell of the heap, and `tail` that points to the last cell. These two cells contain two values
struct Node {
    bool lock;
    int val;
    Node* next;
    bool mark;
}

locate(e):
local p, c
1 while (true)
2 p := Head;
3 c := p.next;
4 while (c.val < e)
5 p := c;
6 c := c.next
7 lock(p); lock(c);
8 if (!p.mark && !c.mark && p.next == c)
9 return (p, c);
10 else
11 unlock(p);
12 unlock(c);

add(e):
local p, c, n, r
1 (p, c) := locate(e);
2 if (c.val != e)
3 n := new Node(
4     false, e, c, false);
5 p.next := n;
6 r := true;
7 else r := false;
8 unlock(p); unlock(c);
9 return r;

rmv(e):
local p, c, n, r
1 (p, c) := locate(e);
2 if (c.val = e)
3 c.mark := true;
4 n := c.next;
5 p.next := n;
6 r := true;
7 else r := false;
8 unlock(p); unlock(c);
9 return r;

cnt(e):
local c
1 c := Head;
2 while (c.val < e)
3 c := c.next
4 b := c.mark
5 if (!b && c.val = e)
6 return true;
7 else
8 return false;

Figure 3.1. Lazy Set Algorithm

that are smaller and larger, respectively, than data values of all cells that may be inserted in the set. The algorithm also contains the subroutine locate that returns a structure containing the cells on either side of e. In more detail, the locate method traverses the list using along two local variables p and c, starting at the head of the list and moves forward the list (line 2), comparing c.val to e. When c is pointed to the first cell whose value of val is greater than or equal to e, the traversal stops, and the method locks cells pointed to by p and c (line 7) so that no other thread can update fields of p and c. Thereafter, if both p.mark and c.mark are false and p.next = c meaning that there is no added cell from other threads between p and c, the method returns the pair (p, c) (line 9). Otherwise, it unlocks cells pointed to by p and c and tries traversing again from the head of the list.

The add(e) method calls locate(e) at line 1 to locate the position in the list where e is to be inserted. Its local variables p and c are assigned the first and second values of the pair returned by locate(e), respectively. If c.val = e, meaning that a cell whose value of val is equal to e is already in the list, the method unlocks p and c and returns false (line 7-8). Otherwise, a new cell n is created (line 3), and inserted into the list by linking it into the list between the p and c pointers returned by locate (line 3-4). Then, the method unlocks cells pointed to by p and c and returns true.
The \texttt{rmv(e)} method also calls \texttt{locate} at line 1 to locate the position in the list where \texttt{e} is to be inserted. If \texttt{c.val \neq e} meaning that a cell with \texttt{val e} is not already in the list, the method unlocks \texttt{p} and \texttt{c} and returns \texttt{false} (line 7-9). Otherwise, cell \texttt{c} is logically removed (line 3) by assigning \texttt{true} to the \texttt{mark} field of \texttt{c}, and unlinking it from the list (line 4-5). Then, the method unlocks cells pointed to by \texttt{p} and \texttt{c} and returns \texttt{true}.

The \texttt{ctn(e)} method traverses the list by using local variable \texttt{c}, ignoring whether nodes are marked or not, until \texttt{c} is set to the first cell with a value of \texttt{val} greater than or equal to \texttt{e}. It simply returns \texttt{true} if and only if the cell pointed to by \texttt{c} is unmarked with the desired value of \texttt{val} equal to \texttt{e}.

3.1 Linearizability

In a concurrent program, the methods of the different executing threads can overlap in time. Thus, a \texttt{rmv} method that executes in parallel with an \texttt{add} method for the same key may or may not find the element in the set, depending on how the individual method statements overlap in time. For a user of the data structure, it is important to know precisely what can happen when several methods access a data structure concurrently, without inspecting the code of each method. Such a user would want to have a criterion for how operations take effect, which considers only the points in time of method calls and returns. The most widely accepted such condition is linearizability. Linearizability defines consistency for the history of call and response events generated by an execution of the program at hand [74]. Intuitively, linearizability requires every method to take effect at some point (linearization point) between its call and return events. A linearization point is intuitively a moment where the effect of the method becomes visible to other threads. An execution of a (concurrent) system is modelled by a (concurrent) history, which is a finite sequence of method call and return events. A (concurrent) history is linearizable if and only if there is some order for the effects of the actions that corresponds to a valid sequential history. The valid sequence history can be generated by an execution of the sequential specification object. A concurrent object is linearizable iff each of its histories is linearizable.

Figure 3.2 provides an example of a trace of method calls of a concurrent program implementing a set. In the trace, each method takes effect instantaneously at it is linearization point between call and return events [74]. When we order methods according to their linearization points, we get a totally ordered sequence that respects the sequential specification of the set.
Figure 3.2. Linearizability, where the linearization points are marked with .
4. Specifying Linearizability

In the previous sections, we described the correctness criterion of linearizability for concurrent data structures. In this section, we specify linearizability in a way that is suitable for automated verification. We separate the problem of specifying linearizability into several ones.

• To specify the sequential semantics of a data structure in a way that is suitable for automated verification.
• To specify placement of linearization points in executions of a concurrent data structure.

For the first, we use the techniques of observers [11]. For the second, we present a new technique, in which methods are equipped with controllers. Controllers specify so-called “linearization policies”, which prescribe how LPs are placed in executions.

4.1 Observers

Data structures are, by nature, infinite-state objects, since they are intended to carry an unbounded number of data elements. For automated verification, it is desirable with specifications that are constructed without explicitly mentioning such infinite objects. This problem is addressed by observers [?]. Observers specify allowed sequences of operations by constraining their projection on a small number of data elements. They exploit the assumption that the data structure handles all data elements in the same way. Observers are finite automata extended with a finite set of observer registers that assume values in the domain of data elements. At initialization, the registers are nondeterministically assigned arbitrary values, which never change during a run of the observer. Transitions are labeled by operations that may be parameterized on registers.

Observers are used as acceptors of sequences of operations on the data structure. The observer processes such sequences one operation at a time. If there is a transition, whose label, after replacing registers by their values, matches the operation, such a transition is performed. If there is no such transition, the observer remains in its current state. The observer accepts a sequence if it can be processed in such a way that an accepting state is reached. The observer is defined in such a way that it accepts precisely those sequences of abstract operations that are not allowed by the semantics of the data structure. We use observers to give exact specifications of the behaviours of data
structures such as sets, queues, and stacks. This is best illustrated by an example. Fig. 4.1 depicts an observer that accepts the sequences of method invocations that are not allowed by a sequential specification set data structure. The observer has three states $s_0$, $s_1$ and $s_2$. It also has one register which carries an arbitrary tracked data value. The initial state $s_0$ corresponds to positions in the runs where the non-deterministically tracked value stored in the observer register $x$ is not present in the set (i.e. each time it has been inserted it got deleted afterwards). The state $s_1$ corresponds to positions in the runs where the tracked value is present in the set (i.e, it has not been deleted since it was last inserted). The accepting state $s_2$ corresponds to positions in the runs where the non-allowed behavior captured by the observer has been observed. The captured bad specification are those where a data value is deleted or found although it is not present in the set, or a data value is not found or cannot be deleted although it is already present in the set.

4.2 Linearization Policies

In order to prove linearizability, the most intuitive approach is to find a linearization point (LP) in the code of the implementation, and show that it is the single point where the effect of the operation takes place [? , 24 , ?]. However, for a large class of linearizable implementations, it is not possible to assign fixed LPs in the code of their methods, but depend on actions of other threads in each particular execution. For example, in the Lazy Set algorithm, a successful $rmv(e)$ method has its LP at line 3, and an unsuccessful $rmv(e)$ has its LP at line 2 when the test $c.val = e$ evaluates to false. A successful $add(e)$ method has its LP at line 4 and an unsuccessful $add$ has its LP at line 2 when the test $c.val != e$ evaluates to false. The successful $ctn$ is linearized at line 4 then the value of $b$ is true. However it is not possible to assign a fixed LP in the code of the $ctn$ method when it is unsuccessful.

To see why unsuccessful $ctn$ method invocations can not have fixed LPs, note that the naive attempt of defining the LP at line 4 provided that the test
\[\text{add}\]
\[\rho_1: \text{when} \bullet \text{provided } pc=4 \text{ emit } \text{add}(e, \text{true}) \text{ broadcast } \text{add}(e)\]
\[\rho_2: \text{when} \bullet \text{provided } pc=2 \& \& (c.\text{val} = e) \text{ emit } \text{add}(e, \text{false})\]
\[\text{rmv}\]
\[\rho_3: \text{when} \bullet \text{provided } pc=3 \text{ emit } \text{rmv}(e, \text{true})\]
\[\rho_4: \text{when} \bullet \text{provided } pc=2 \& \& c.\text{val} \neq e \text{ emit } \text{rmv}(e, \text{false})\]
\[\text{ctn}\]
\[\rho_5: \text{when} \bullet \text{provided } pc=4 \& \& !b \& \& c.\text{val} = e \text{ emit } \text{ctn}(e, \text{true})\]
\[\rho_6: \text{from } q_0 \text{ when} \bullet \text{provided } pc=4 \& \& (!b \& \& c.\text{val}=e) \text{ emit } \text{ctn}(e, \text{false}) \text{ goto } q_0\]
\[\rho_7: \text{from } q_0 \text{ when} (\text{add}(e), b) \text{ provided } 1\leq pc<4 \text{ emit } \text{ctn}(e, \text{false}) \text{ goto } q_1\]

Figure 4.2. Reaction Rules for Controllers of Lazy Set.

at line 5 fails will not work. Namely, the \text{ctn} method may traverse the list and arrive at line 4 in a situation where the element \(e\) is not in the list (either \(e\) is not in any cell, or the cell containing \(e\) is marked). However, before executing the command at line 4, another thread performs an add operation, inserting a new cell containing the element \(e\) into the list. The problem is now that the \text{ctn} method cannot “see” the new cell, since it is unreachable from the cell currently pointed to by the variable \(b\) of the \text{ctn} method. If the \text{ctn} method would now try to linearize an unsuccessful \text{ctn}, this would violate the semantics of a set, since the add method just linearized a successful insertion of \(e\).

There have been several previous works dealing with the problems of non-fixed linearization points [129, 43, 44, 113, 48, 112, 121, 37]. However, they are either manual approaches without tool implementation or not general enough to cover various types of concurrent programs. In this thesis we handle non-fixed linearization points by providing a mechanism for assigning LPs to executions, which we call linearization policies.

A linearization policy is expressed by associating method invocations to a controller, which is responsible for generating operations announcing the occurrence of LPs during each method invocation. The controller is occasionally activated, either by its thread or by another controller, and mediates the interaction of the thread with the observer as well as with other threads.

To add controllers, we first declare some statements in each method to be triggering: these are marked by the symbol \(\bullet\) as in Figure 3.1. We specify the behaviour of the controller, belonging to a method \(m\), by a set of reaction rules. To define these rules, we first define different types of events that are used to specify their behaviours. Note that an operation is of the form \(m(d_{\text{in}}, d_{\text{out}})\) where \(m\) is a method name and \(d_{\text{in}}, d_{\text{out}}\) are data values. Operations are emitted by the controller to the observer to notify that the thread executing the method performs a linearization of the corresponding method with the given input and output values. Next, we fix a set \(\Sigma\) of broadcast messages, each with a fixed arity, which are used for synchronization between controllers. A broadcast
message is formed by supplying data values as parameters. In reaction rules, these data values are denoted by expressions over the variables of the method, which are evaluated in the current state when the rule is invoked. We define two types of reaction rules:

- A triggered rule, of form `when • provided cnd emit op broadcast se`, specifies that whenever the method executes a triggering statement and the condition `cnd` evaluates to `true`, then the controller performs a reaction in which it emits the operation obtained by evaluating `op` to the observer, and broadcasts the message obtained by evaluating `se` to the controllers of other threads. The broadcast message `se` is optional.

- A receiving rule, of form `when ⟨re, ord⟩ provided cnd emit op`, specifies that whenever the observer of some other thread broadcasts the message obtained by evaluating `re`, and `cnd` evaluates to `true`, then the controller performs a reaction where it emits the operation obtained by evaluating `op` to the observer. Note that no further broadcasting is performed. The interaction of the thread with the observer may occur either before or after the sender thread send the broadcast, according to the flag `ord`.

A controller may also use a finite set of states, which restrict the possible sequences of reactions by a controller in the standard way. Whenever such states are used, the rule includes source and target states using keywords `from` and `goto`. In Figure 4.2, the rule ρ7 changes the state from q0 to q1, meaning that no further applications of rules ρ6 or ρ7 are possible, since they both start from state q0. Rules that do not mention states can be applied regardless of the controller state and leave it unchanged.

Let us illustrate how the reaction rules for controllers in Figure 4.2 specify LPs for the algorithm in Figure 3.1. Here, a successful rmv method has its LP at line 3, and an unsuccessful rmv has its LP at line 2 when the test `c.val = e` evaluates to `false`. Therefore, both these statements are marked as triggering. The controller has a reaction rule for each of these cases: in Figure 4.2: rule ρ3 corresponds to a successful rmv, whereas rule ρ4 corresponds to an unsuccessful rmv. Rule ρ4 states that whenever the rmv method executes a triggering statement, from a state where `pc=2` and `c.val != e`, then the operation `rmv(e,false)` will be emitted to the observer.

A successful add method has its LP at line 4. Therefore, the controller for add has the triggered rule ρ1 which emits the operation `add(e,true)` to the observer. In addition, the controller also broadcasts the message `add(e)`, which is received by any controller for a ctn method which has not yet passed line 4, thereby linearizing an unsuccessful ctn(e) method by emitting `ctn(e,false)` to the observer. The keyword `b` denotes that the operation `ctn(e,false)` will be presented before `add(e,true)` to the observer. Since the reception of `add(e)` is performed in the same atomic step as the triggering statement at line 4 of the add method, this describes a linearization pattern, where a ctn method, which has not yet reached line 4, linearizes an unsuc-
cessful \texttt{cnt}-invocation just before some other thread linearizes a successful add of the same element.

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure4.3.png}
\caption{An example of linearization policies of Lazy Set algorithm}
\end{figure}

Let us describe an example of how the reaction rules in fig 4.3 handles non-fixed linearization points of unsuccessful \texttt{cnt}(2) method. Figure 4.3 gives an example of Lazy Set with three threads $T_1$, $T_2$, and $T_3$. The thread $T_1$ is executing the \texttt{add}(2) method to insert the cell whose value of $val$ is equal to 2 into the set, while both $T_2$ and $T_3$ are executing \texttt{cnt}(2) to search for a cell whose value of $val$ is equal to 2. When thread $T_1$ reaches the triggering statement at line 4 of the \texttt{add}(2) method at the step $T_1$, the controller rule $\rho_1$ is activated to inform the observer that an add operation with argument 2 has been performed, and that the outcome of the operation is $\text{true}$ (the operation was successful) in step 6. However, before that, the controller will help other threads to linearize. This is done by broadcasting a message \texttt{add}(2) to the threads $T_2$, $T_3$ which are executing \texttt{cnt}(2) in steps 2. When these threads get the message then the rule $\rho_7$ is activated to inform the observer that the element 2 is not in the list in threads $T_2$, $T_3$ respectively, in steps 3 and 5.

\textit{Verifying Linearization Policies}

By using an observer to specify the sequential semantics of the data structure, and defining controllers that specify the linearization policy, the verification of linearizability is reduced to establishing four conditions: [(i)]

\begin{itemize}
\item each method invocation generates a non-empty sequence of operations,
\item the last operation of a method conforms to its parameters and return value,
\end{itemize}
only the last operation of a method may change the state of the observer, and
the sequence of all operations cannot drive the observer to an accepting state.
Our verification framework automatically reduces the establishment of these
conditions to a problem of checking control state reachability. This is done
by augmenting the observer by a monitor. The monitor is automatically gen-
erated. It keeps track of the state of the observer, and records the sequence
of operations and call and return actions generated by the threads. For each
thread, it keeps track of whether it has linearized, whether it has caused a state
change in the observer, and the parameters used in its last linearization. Us-
ing this information, it goes to an error state whenever any of the above four
conditions is violated.
5. Shape Analysis

Pointers and heap-allocated storage are features of all modern imperative programming languages. They are among the most complicated features of imperative programming language: updating pointer variables (or pointer-fields of records) may have large side effects. For example, dereferencing a pointer that has been freed will lead to segmentation fault in a C or C++ program. Such side effects also make program analysis harder, because they make it difficult to compute aliasing relationships between different pointers in a program. Aliasing arises when the same memory location can be accessed using different names. For instance, consider the instruction statement \( x.f := y \) in an imperative language, where \( x \) and \( y \) are pointer variables. Its effect is to assign the value of the pointer \( y \) to the cell pointed to by \( x.f \). In order to update aliasing information of \( y \), we have to require information about all the cell pointed by \( x.f \) which is not an easy task.

In verification and program analysis, it is a problem to deduce and describe how the heap-allocated memory is organized. E.g., program invariants must often describe how the heap-allocated memory is structured in order to infer the effects of statements that dereference pointer fields. This is the topic of “shape analysis”.

5.1 Previous Approaches

Shape analysis is a generic term denoting static program-analysis techniques that attempt to discover and verify properties of the heap contents in (usually imperative) programs. The shape analysis problem becomes more challenging in concurrent programs that manipulate pointers and dynamically allocated objects, which are usually complicated. In concurrent programs, different threads interact in complex ways, which are difficult to foresee in the analysis. Several approaches for representing the possible structures of the heap have been proposed. TVLA (Three-Valued Logic Analysis) [109] is one of the first and one of the most popular shape analysis methods. It is based on a three-valued first-order predicate logic with transitive closure. Intuitively, a concrete heap structure is represented by a finite set of abstract summary cells, each of them representing a set of concrete cells. Summary cells are obtained by merging several heap cells that agree on the values of a chosen set of unary abstraction predicates. An important aspect of TVLA is that it automatically
Figure 5.1. A heap configuration of the Lazy Set Algorithm where head and tail are global variable, and p and c are two local variables. The symbols ✔ and ✗ represent the Boolean values true and false.

generates the abstract transformers from the concrete semantics; these transformers are guaranteed to be sound, and precise enough to verify wide ranges of applications. However, it cannot fully automatically handle all programs, and one may have to extend it by appropriate predicates. By synthesising appropriate predicates TVLA is able to express the invariants in the program. This problem is even more difficult with complicated heap structures such as skip-lists, trees, or arrays of lists.

There are several other approaches based on the use of logics to present heap configurations. The logics can be separation logic [106, 87, 23, 126, 52, 39, 84, 101, 56], monadic second-order logic [97, 77, 86] and others [108, 127]. Among these works, the works based on separation logic are more efficient than the others. The reason for that is that their approaches effectively decompose the heap into disjoint components and treat them independently. However, most of the techniques based on separation logic are either specialised for some particular data structure, or they need to be provided inductive definitions of the data structures. In addition, their entailment checking procedures are either for specific classes of data structures or based on folding/unfolding inductive predicates in the formulae and trying to obtain a syntactic proof of the entailment.

The above problems in shape analysis can be addressed by automata-based techniques using the generality of the automata-based representation. Finite tree automata, for instance, have been shown to provide a good balance between efficiency and expressiveness. The work [33] uses a finite tree automaton to describe a set of tree parts and represent non-tree edges of heaps by using regular “routing” expressions. Finite tree transducers are used to compute sets of reachable configurations, and symbolic configurations are abstracted collapsing certain states of the automata. The refinement technique called counterexample-guided abstraction refinement (CEGAR) technique is used during the run of the analysis. This technique is fully automated and can handle complex data structures such as binary trees with linked leaves. However, it suffers from inefficiency and it also cannot handle concurrent programs.
Figure 5.2. The forest representation of the heap configuration in Fig. 5.1.

The problem of inefficiency of the previous technique can be solved by the approach based on forest automata \[?\]. In this representation, a heap is split in a canonical way into several tree components whose roots are the so-called cut-points. Cut-points are either cells which are pointed to by program variables or have several incoming edges. The tree components can refer to the roots of each other, and hence they are “separated” much like heaps described by formulae joined by the separating conjunction in separation logic \[106\]. Using this decomposition, sets of heaps with a bounded number of cut-points can then be represented by so called forest automata (FA). An FA is a tuple of TA, where each of the tree automata within the tuple accepts trees whose leaves may refer back to the roots of any of these trees. A forest automaton then represents exactly the set of heaps that may be obtained by taking a single tree from the language of each of the component tree automata and by gluing the roots of the trees with the leaves referring to them. Moreover, they allow alphabets of FA to contain nested FA, leading to a hierarchical encoding of heaps, allowing us to represent even sets of heaps with an unbounded number of cut-points (e.g., sets of DLL, skiplists).

Let us take an example of how to split a heap into small tree components. Figure 5.2 shows five tree components obtained by splitting the heap in figure 5.1. These components are named as 1, 2, 3, 4 and 5 from left to right. Each root of a tree component is a cut-point in the heap in figure 5.1. These cut-points are cells pointed to by variables head, tail, p, c and the cell which has two incoming pointers. In each tree component, the red node shows to which tree component it refers. For instance, the tree component 1 refers to tree component 2, and both tree components 2 and 3 refer to tree component 4.

This forest automata approach is fully automatic and able to verify various classes of data structures, including complicated structures such as trees and skip-lists with a bounded number of levels. However, the approach can not verify properties related to data values of heap cells such as sortedness in the
lazy set algorithm. Therefore, in this thesis, we extend their work to verify data properties.

The last approach that we will mention is based on graph grammars describing heap graphs [69, 68]. The approach is to model heap states as hypergraphs, and represent both pointer operations and abstraction mappings by hypergraph transformations. The presented approaches differ in their degree of specialisation for a particular class of data structures, their efficiency, and their level of dependence on user assistance (such as definition of loop invariants or inductive predicates for the considered data structures).

5.2 Our Approaches

In this thesis, we propose three approaches for symbolic representation of heap. In paper I, we propose a novel approach of extending the forest automata approach [7] by expressing relationships between data elements associated with cells of the heap by two classes of constraints.

Figure 5.3. A graph and its forest representation

- Local data constraints are associated with transitions of each TA and capture relationships between data of neighboring cells in a heap; they can be used, e.g., to represent ordering internal to data structures such as sorted linked lists and binary search trees.
- Global data constraints are associated with states of TA and capture relationships between data in distant parts of the heap. Intuitively, a global data constraint between two TAs captures the data relationship between cells of two heaps accepted by these two TAs.
Figure 5.3 shows an example of how to represent the heap in 5.1 by a set of tree automata with added data constraints. In the figure, the local data constraints are located along the solid arrows between cells, whereas global constraints are located along the blue arrows. The global constraint $\prec_{aa}$ means that data values of all cells in the left hand side are smaller than data values of all cells in the right hand side. The local constraint $\prec_{rr}$ means that the data value of the left hand side cell is smaller than data value of the cell in the right hand side. For instance, in tree component 1, the data value of the cell pointed to by head is smaller than its successor whose data value is 4, and data values of all cells in tree component 1 are smaller than data values of all cells in tree component 2. We just show here small examples of data constraints; details about different types of constraints can be found in paper I.

This approach was applied to verification of sequential heap manipulation programs. The approach is general and able to verify properties of many types of sequential programs without any manual step. However, due to the complexity of tree automata operations, this approach is not optimal to handle concurrent programs where a large number of states and computation are needed. In order to verify concurrent data structures with an unbounded number of threads, the thread-modular reasoning [58] is a promising approach for this challenge. Its high efficiency is achieved by abstracting the interaction between threads. The approach verifies a concurrent data structure by generating an invariant that correlates the global state with the local state of an arbitrary thread. In other words, it only keeps track of the shape as viewed by one thread, while abstracting away all the other threads. The thread-modular approach includes a step which takes the information about one thread and intersects it with the information from another thread, in order to take into account the interference of all the other threads on the first thread. The forest automata approach is not suitable for this thread-modular approach. The reason is that computing intersection between FAs is not efficient in concurrent systems where the number of FAs is huge.

Therefore, in paper II, we adapt FA to the new setting by providing a symbolic encoding of the heap structure, which is less precise than the forest automata approach in paper I. However it is precise enough to allow the verification of concurrent data structures, and efficient enough to make the verification procedure feasible in practice.

The main idea of the abstraction is to have a more precise description of the parts of the heap, called heap segments that are visible (reachable) from global variables, and to make a succinct representation of the parts that are local to the threads. Intuitively, a heap segment can be characterized by a TA with data constraints. More concretely, we will extract a set of heap segments between two cut-points which are analogous to cut-points in the forest automata approach. For each segment, we will store a summary of the content of the heap along the segment. This summary consists of three parts where each part contains different pieces of information, including
• the values of data fields of cells along the segment which have finite domains, and
• the ordering among data values of fields of cells along the segment which have integer values.
• The sequence of observer registers that appear in the segment.

Figure 5.4. A symbolic representation of the configuration of lazy set in Figure 5.1

Figure 5.4 gives our symbolic abstraction of the heap in figure 5.1 where the observer register \( z \) is equal to 3. In this figure, there are four segments obtained from five cut-points. In each segment, the red box contains ordering information between data values of cells along the segment, the green box contains information about the values of fields \( \text{mark} \) and \( \text{lock} \) of cells along the segment. Finally, the blue box contains the information about the sequence of observer registers. In this example, let us consider the first segment between the two cells pointed to by \( \text{head} \) and \( p \). The sequence of observer registers is \( z^1 \), it means that between two cut-points there is exactly one cell whose value of the \( \text{val} \) field is equal to the observer register \( z \).

This approach is efficient but it is not optimal for complicated concurrent data structures like trees, lists of lists or skiplists. The approach is therefore specialized for SLLs.

In paper III, we present an approach which can handle concurrent programs implemented using both simple and complex data structures. More precisely, we can handle well-known data structures like singly-linked lists, skiplists, and lists of lists.

**Heap abstraction for singly-linked lists**
The main idea of the approach is to represent a set of heap states by a set of fragments. A fragment represents two heap cells that are connected by a pointer field. For each of its cells, the fragment represents the contents of its non-pointer fields, together with information about how the cell can be reached from the program’s pointer variables, under the applied data abstraction. The fragment contains both
• local information about the cell’s fields and variables that point to it, as well as
• global information, representing how each cell in the pair can reach to and be reached from (by following a chain of pointers) a small set of globally significant heap cells.

A set of fragments represent the set of heap structures in which each pair of pointer-connected nodes is represented by some fragment in the set. Put differently, a set of fragments describes the set of heaps that can be formed by “piecing together” pairs of pointer-connected nodes that are represented by some fragment in the set. This “piecing together” must be both locally consistent (appending only fragments that agree on their common node), and globally consistent (respecting the global reachability information). Figure 5.5 shows a set of fragments that is sufficient to represent the configuration in Figure 5.1. There are 7 fragments, named \( v_1, \ldots, v_7 \). Fragment \( v_7 \) consists of a tag that points to \( \perp \). All other fragments consist of a pair of pointer-connected tags.

**Heap abstraction for skiplists**

Let us illustrate how pairs of heap cells can be represented by fragments. Before going detail to the example, let us describe in short the definition of skiplist. A skiplist consists of a collection of sorted linked lists, each of which is located at a *level*, ranging from 1 up to a maximum value. Each skiplist node has a key value and participates in the lists at levels 1 up to its *height*. The skiplist has sentinel head and tail nodes with maximum heights and key values \(-\infty\) and \(+\infty\), respectively. The lowest-level list (at level 1) constitutes an ordered list of all nodes in the skiplist. Higher-level lists are increasingly
sparse sublists of the lowest-level list, and serve as shortcuts into lower-level lists.

Figure 5.6 shows an example heap state of the skiplist algorithm with three levels. Each heap cell is shown with the values of its mark and key fields. Let us illustrate how pairs of heap nodes can be represented by fragments. Figure 5.7 shows a set of fragments that is sufficient to represents the configuration in Figure 5.6. There are 11 fragments, named $v_1, \ldots, v_{11}$. Three of these ($v_9, v_{10}$ and $v_{11}$) consist of a tag that points to $\bot$. All other fragments consist of a pair of pointer-connected tags. The fragments $v_1, v_2, v_3, v_4, v_5, v_9$ and $v_{10}$ are level-1-fragments, whereas the other fragments are higher level-fragments. The mark field of the input tag of $v_7$ is true, whereas the mark field of other tags of other fragments are false. The two left-most cells in Figure 5.6 are represent by the level 1-fragment $v_1$ in Figure 5.7. Here, the variable $\text{preds}[3]$ is represented by $\text{preds[higher]}$. The mapping $\pi_1$ represents the data abstraction of the key field, here saying that it is smaller than the value 9 of the observer register. Note that, in our approach, the data abstraction is a mapping function from data value to a set of observer registers. The two left-most cells are also represented by a higher-level fragment, viz. $v_6$. The pair consisting of the two sentinel cells (with keys $-\infty$ and $+\infty$) are represented by the higher-level fragment $v_7$. In each fragment, the abstraction values of non-pointer fields are shown represented inside each tag of the fragment. The data relation is shown as a label on the arrow between two tags. Above each tag is pointer variables. The first row under each tag is reached from information, whereas the second row is reached to information.
Figure 5.7. Fragment abstraction of skiplist algorithm
6. Related Work

This chapter reviews related work, along the main topics of this thesis including: verification of Linearizability of concurrent algorithms, shape analysis of sequential and concurrent programs.

6.1 Verification of Linearizability

There are several previous work on verification of linearizability of concurrent programs, see eg the survey [50]. Since the definition of linearizability by Herlihy and Wing [75] the most common approaches of conformance to a simple abstract specification use refinement techniques, which establish simulation relations between the implementation and specification, using partly manual techniques. Vafeiadis [119] uses forward and backward simulation relations together with history or prophecy variables to prove linearizability. The simulation techniques are mostly manual, but they have also been mechanized. Mechanical proofs of linearizability, using interactive theorem provers, have been reported in [43, 48, 113, 112]. Colvin et al. [43] verify the lazy set algorithm in PVS, using a combination of forward and backward simulations.

There are several works on automatic verification of linearizability. the works [11, 24, 120] all require fixed linearization points which are manually annotated by users. In [121], Vafeiadis develops an automatic technique for proving linearizability that employs instrumentation to verify logically pure executions. This work can also handle non-fixed LPs, but only for read-only methods, i.e, methods that do not modify the heap. This means that the method cannot handle algorithms like the Elimination queue [96], HSY stack [71], CCAS [66], RDCSS [66] and HM set [73] that we consider in the thesis. In addition, their shape abstraction is not powerful enough to handle algorithms like Harris set [65] and Michael set [94] that are also handled by techniques in this thesis.

Chakraborty et al. [72] describe an “aspect-oriented” method for modular verification of concurrent queues that they use to prove linearizability of the Herlihy/Wing queue. They define a set of constraints on the ordering between method calls and returns that characterise the behaviour of a queue and show that Herlihy/Wing’s queue satisfies these constraints. Bouajjani et al. [31] extended this work to show that verifying linearizability for certain fixed abstract data types, including queues and stacks, is reducible to control-state reachability. We can incorporate this technique into our framework by a suitable
construction of observers. We have not yet been able to apply this method to sets.

Vechev et al. [123] check linearizability with user-specified non-fixed LPs, using a tool for finite-state verification. Their method assumes a bounded number of threads, and they report state space explosion when having more than two threads. Dragoi et al. [51] describe a method for proving linearizability that is applicable to algorithms with non-fixed LPs. However, their method needs to rewrite the implementation so that all operations have linearization points within the rewritten code. Černý et al. [36] show decidability of a class of programs with a bounded number of threads operating on concurrent data structures. Their technique can be used to verify optimistic implementations of concurrent set data structures with the help of programmer annotations.

In [100], O’Hearn et al. define a hindsight lemma that provides a non-constructive evidence for linearizability. The lemma assumes that the algorithm maintains certain simple invariants which are resilient to interference, and which can themselves be verified using purely thread-local proofs. The lemma is used to prove linearizability of an optimistic variant of the lazy set algorithm. Based on this lemma the work of Zhu et al. [129] describe a tool combining the hindsight lemma and an SMT solver to prove linearizability; they can handle several specific set, queue, and stack algorithms. For queue algorithms, their technique can handle queues with helping mechanism except for HW queue [74] which is handled in this thesis. For set algorithms, the authors can only handle those that perform an optimistic contains (or lookup) operation by applying the hindsight lemma from [100]. Hindsight-based proofs focus on a collection of invariants specific to the traversals they study, algorithms with non-optimistic contains (or lookup) operation like HM [73], Harris [65] and Michael [94] sets cannot be verified by their technique.

In [92, 93], they present a concurrent separation logic for automating linearizability proofs for concurrency library implementations. Moreover, their approaches are not yet fully automated: the user provides an invariant describing the properties of each node comprising the data structure in the shared heap.

We have not found any report in the literature of a verification method that is sufficiently powerful to automatically verify the class of concurrent set implementations based or sorted and non-sorted singly-linked lists having non-optimistic contains (or lookup) operations we consider. For instance

the lock-free sets of HM [73], Harris [65], or Michael [94], or the unordered set of [128],
6.2 Shape Analysis

6.2.1 Shape Analysis for Sequential Programs

Our approach builds on the fully automated FA-based approach for shape analysis of programs with complex dynamic linked data structures [64, 13]. We significantly extend this approach by allowing it to track ordering relations between data values stored inside dynamic linked data structures.

For shape analysis, many other formalisms than FA have been used, including, e.g., separation logic and various related graph formalisms [126, 53, 38, 52], other logics [109, 77], automata [34], or graph grammars [69]. Compared with FA, these approaches typically handle less general heap structures (often restricted to various classes of lists) [126, 52], they are less automated (requiring the user to specify loop invariants or at least inductive definitions of the involved data structures) [53, 38, 52, 69], or less scalable [34].

Verification of properties depending on the ordering of data stored in SLLs was considered in [29], which translates programs with SLLs to counter automata. A subsequent analysis of these automata allows one to prove memory safety, sortedness, and termination for the original programs. The work is, however, strongly limited to SLLs. In our technique, we get inspired by the technique that [29] uses for dealing with ordering relations on data, but we significantly redesign it to be able to track not only ordering between simple list segments but also general heap shapes described by FA. In order to achieve this, we had to not only propose a suitable way of combining ordering relations with FA, but we also had to significantly modify many of the operations used over FA.

In [2], another approach for verifying data-dependent properties of programs with lists was proposed. However, even this approach is strongly limited to SLLs, and it is also much less efficient than our current approach.

Verification of properties of programs depending on the data stored in dynamic linked data structures was considered in the context of the TVLA tool [78] as well. This approach assumes a fixed set of shape predicates such as pointer equality, tests for nil pointers, and garbage cells, as well as reachability along pointers. It uses inductive logic programming to learn predicates needed for tracking non-pointer data. The experiments presented in [78] involve verification of sorting of several programs on SLLs (merging, reversal, bubble-sort, insert-sort) as well as insertion and deletion in BSTs. Our approach is more efficient. Moreover, for BSTs, we verify that a node is greater/smaller than all the nodes in its left/right subtrees (not just than the immediate successors as in [78]).

An approach based on separation logic extended with constraints on the data stored inside dynamic linked data structures and capable of handling size, ordering, as well as bag properties was presented in [38]. Using the approach, various programs with SLLs, DLLs, and also AVL trees and red-black trees were verified. The approach, however, requires the user to manually provide
inductive shape predicates as well as loop invariants. Later, the need to provide loop invariants was avoided in [104], but a need to manually provide inductive shape predicates remains.

Another work that targets verification of programs with dynamic linked data structures, including properties depending on the data stored in them, is [126]. It generates verification conditions in an undecidable fragment of higher-order logic and discharges them using decision procedures, first-order theorem proving, and interactive theorem proving. To generate the verification conditions, loop invariants are needed. These can either be provided manually or sometimes synthesized semi-automatically using the approach of [109]. The latter approach was successfully applied to several programs with SLLs, DLLs, trees, trees with parent pointers, and 2-level skip lists. However, for some of them, the user still had to provide some of the needed abstraction predicates. Several works, including [30], define frameworks for reasoning about pre- and post-conditions of programs with SLLs and data. Decidable fragments, which can express more complex properties on data than we consider, are identified, but the approach does not perform fully automated verification, only checking of pre-post condition pairs.

6.2.2 Shape Analysis for Concurrent Programs

There are several approaches for automated verification of concurrent algorithms that are limited to the case of singly-linked lists [11, 67, 17, 114, 121]. Furthermore, many of these techniques impose additional restrictions on the considered verification problem, such as bounding the number of accessing threads [19, 123, 36].

In [11], concurrent programs operating on SLLs are analyzed using an adaptation of a transitive closure logic [26], combined with tracking of simple sortedness properties between data elements; the approach does not allow to represent patterns observed by threads when following sequences of pointers inside the heap, and so has not been applied to concurrent set implementations. In paper II, we extended this approach to handle SLL implementations of concurrent sets by adapting a well-known abstraction of singly-linked lists [90] for concurrent programs. The resulting technique is specifically tailored for singly-links. Our fragment abstraction in paper III is significantly simpler conceptually, and can therefore be adapted also for other classes of heap structures. The approach of paper II is the only one with a shape representation strong enough to verify concurrent set implementations based on sorted and non-sorted singly-linked lists having non-optimistic contains (or lookup) operations we consider, such as the lock-free sets of HM [73], Harris [65], or Michael [94], or unordered set of [128]. Our fragment abstraction can handle them as well as also algorithms employing skiplists and arrays of singly-linked lists.
There is no previous work on automated verification of skiplist-based concurrent algorithms. The work [110] generates verification conditions for statements in sequential skiplist implementations. All these works assume that skiplists have the well-formedness property that any higher-level list is a sublist of any lower-level list, which is true for sequential skiplist algorithms, but false for several concurrent ones, such as [73, 85].

Concurrent algorithms based on arrays of SLLs, and including timestamps, e.g., for verifying the algorithms in [49], have shown to be rather challenging. Only recently has the TS stack been verified by non-automated techniques [32] using a non-trivial extension of forward simulation, and the TS queue been verified manually by a new technique based on partial orders [81, ?]. We have verified both these algorithms automatically using fragment abstraction.

Our fragment abstraction is related in spirit to other formalisms that abstract dynamic graph structures by defining some form of equivalence on its nodes (e.g., [124, 109]). These have been applied to verify functional correctness fine-grained concurrent algorithms for a limited number of SLL-based algorithms. Fragment abstraction’s representation of both local and global information allows to extend the applicability of this class of techniques.
7. Summary of Contributions

In this chapter, we give a short overview of our three peer-reviewed papers. We will explain the problem addressed by each paper and its main contributions.

7.1 Paper I: Verification of Heap Manipulating Programs with Ordered Data by Extended Forest Automata

Automated verification of programs that manipulate complex dynamic linked data structures is a challenging problem in software verification. The problem becomes even more challenging when program correctness depends on relationships between data values that are stored in the dynamically allocated structures.

In this paper, we present a general framework for verifying such programs. The underlying formalism of our framework is that of forest automata (FA) which has previously been developed for representing sets of reachable configurations of programs with complex dynamic linked data structures but without data stored in nodes. In the FA framework, a heap graph is represented as a composition of tree components. Sets of trees can be represented by tree automata, sets of heap graphs can then be represented by tuples of tree automata (TA). We extend FA by adding constraints between data elements associated with nodes of the heaps represented by FA, and we present extended versions of all operations needed for using the extended FA in a fully-automated verification approach, based on abstract interpretation. Technically, we express relationships between data elements associated with nodes of the heap graph by two classes of constraints. Local data constraints are associated with transitions of TA and capture relationships between data of neighbouring nodes in a heap graph; they can be used, e.g., to represent ordering internal to some structure such as a binary search tree. Global data constraints are associated with states of TA and capture relationships between data in distant parts of the heap. In order to obtain a powerful analysis based on such extended FA, the entire analysis machinery must have been redesigned, including a need to develop mechanisms for propagating data constraints through FA, to develop a new inclusion check between extended FAs, and to define extended abstract transformers.
The resulting method allows for verification of pointer programs where the needed inductive invariants combine complex shape properties with constraints over stored data, such as sortedness. The method is fully automatic, quite general, and its efficiency is comparable with other state-of-the-art analyses even though they handle less general classes of programs and/or are less automated.

We have implemented our approach as an extension of the Forester tool and successfully applied it to a number of programs dealing with data structures such as various forms of singly- and doubly-linked lists, binary search trees, as well as skip lists. We presented experimental results from verifying programs dealing with variants of (ordered) lists and trees. To the best of our knowledge, our method is the first one to cope fully automatically with a full C implementation of a 3-level skip list.

7.2 Paper II: Automated Verification of Linearization Policies

Data structures that can be accessed concurrently by many parallel threads are a central component of many software applications, and are implemented in several widely used libraries (e.g., java.util.concurrent). Linearizability is the standard correctness criterion for such concurrent data structure implementations. It states that each operation on the data structure can be considered as being performed atomically at some point, called the linearization point (LP), between its invocation and return. This allows client threads to understand the data structure in terms of atomic actions, without considering the complications of concurrency.

In this paper, we present a novel uniform framework for automatically verifying linearizability of concurrent data structure implementations, which handles the following challenges: Firstly, We handle non-fixed LPs by a novel formalism for specifying linearization policies, by means of so-called controllers. Linearization policies are often described informally by the algorithm designers together with each new data structure implementation when explaining why it is linearizable. Our controllers offer a way to express such policies in a simple and uniform manner. They can express complex patterns for linearization that are much more general than fixed LPs. In detail, a controller is responsible for generating operations announcing the occurrence of LPs during each method invocation. The controller is occasionally activated, either by its thread or by another controller, and mediates the interaction of the thread with the observer as well as with other threads. Secondly, we handle the challenge of an unbounded number of threads by extending the successful thread-modular approach which verifies a concurrent program by generating an invariant that correlates the global state with the local state of an arbitrary thread. Finally, we present a novel symbolic representation of singly-linked
heap structures. We have implemented our technique in a tool, and applied it to specify and automatically verify linearizability of all the implementations of concurrent set, queue, and stack algorithms known to us in the literature, as well as some algorithms for implementing atomic memory read/write operations. To use the tool, the user needs to provide the code of the algorithm together with the controllers that specify linearization policies. To our knowledge, this is the first time all these examples are verified fully automatically in the same framework.

7.3 Paper III: Fragment Abstraction for Concurrent Shape Analysis

A major challenge in automated verification is to develop techniques that are able to reason about fine-grained concurrent algorithms that consist of an unbounded number of concurrent threads, which operate on an unbounded domain of data values, and use unbounded dynamically allocated memory. Existing automated techniques consider the case where shared data is organized into singly-linked lists.

In this paper, we present a technique for automatic verification of concurrent data structure implementations that operate on dynamically allocated heap structures which are more complex than just singly-linked lists. Our approach is the first framework that can automatically verify concurrent data structure implementations that employ singly linked lists, skiplists [15, 23, 39], as well as arrays of singly linked lists [11], at the same time as handling an unbounded number of concurrent threads, an unbounded domain of data values (including timestamps), and an unbounded shared heap.

Our technique is based on a novel shape abstraction, called fragment abstraction, which in a simple and uniform way is able to represent several different classes of unbounded heap structures. Its main idea is to represent a set of heap states by a set of fragments. A fragment represents two heap cells that are connected by a pointer field. For each of its cells, the fragment represents the contents of its non-pointer fields, together with information about how the cell can be reached from the program’s pointer variables. The latter information consists of both: (i) local information, saying which pointer variables point directly to them, and (ii) global information, saying how the cell can reach to and be reached from (by following chains of pointers) other heap cells that are significant from a global perspective, typically since they are pointed to by global variables. A set of fragments represents the set of heap states in which any two pointer-connected nodes is represented by some fragment in the set. Thus, a set of fragments describes the set of heaps that can be formed by “pieced together” fragments in the set. The combination of local and global information in fragments supports precise reasoning about the sequence of cells that can be accessed by threads that traverse the heap by
following pointer fields in cells and pointer variables: the local information captures properties of the cell fields that can be accessed as a thread dereferences a pointer variable or a pointer field; the global information also captures whether certain significant accesses will at all be possible by following a sequence of pointer fields. This support for reasoning about patterns of cell accesses enables automated verification of reachability and other functional properties. Fragment abstraction can (and should) be combined, in a natural way, with data abstractions for handling unbounded data domains and with thread abstractions for handling an unbounded number of threads. For the latter we adapt the successful threadmodular approach [5], which represents the local state of a single, but arbitrary thread, together with the part of the global state and heap that is accessible to that thread. Our combination of fragment abstraction, thread abstraction, and data abstraction results in a finite abstract domain, thereby guaranteeing termination of our analysis. We have implemented our approach and applied it to automatically verify correctness, in the sense of linearizability, of a large number of concurrent data structure algorithms, described in a C-like language. More specifically, we have automatically verified linearizability of most linearizable concurrent implementations of sets, stacks, and queues, and priority queues, which employ singly-linked lists, skiplists, or arrays of timestamped singly-linked lists, which are known to us in the literature on concurrent data structures. For this verification, we specify linearizability using the simple and powerful technique of observers [1, 7, 21, 3], which reduces the criterion of linearizability to a simple reachability property. To verify implementations of stacks and queues, the application of observers can be done completely automatically without any manual steps, whereas for implementations of sets, the verification relies on linearization policies.
8. Conclusions and Directions for Future Work

We have presented, in this thesis, approaches to verify the complex problem of both sequential and concurrent heap manipulating programs. Such programs induce an infinite-state space in several dimensions: they (i) consist of an unbounded number of concurrent threads, (ii) use unbounded dynamically allocated memory, (iii) the domain of data values is unbounded, and (iv) consist of un bounded number of pointers. In addition, the linearization points of some programs depend on the future executions of these programs. In this thesis, we focus on proving both safety properties, and linearization properties for the system. In order to prove safety properties, we define an abstract model of the program, and we employ approximation techniques that ignore irrelevant information so that we can reduce the problem into a finite-state model. In fact, we use an over-approximation, such that the abstract model cover all the behaviours of the original system. However, it might cover other behaviours which are not in the original system. If the bad states are not reached during the computation of reachable states of the abstract model, then the abstract model is considered safe, and so is the original system is also safe. If the bad state is reachable, we have to refine the abstraction. In order to verify linearization properties of a program, we add a specification which expresses its a data structure, using the technique of observers. In our approaches, the user has to provide linearization policies which specify how the program is linearized. We use a technique call controller to specify linearization policies. We then verify that in any concurrent execution of a collection of method calls, the sequence of announced operations satisfies the semantics of the data structure. This check is performed by an observer, which monitors the sequence of announced operations. This reduces the problem of checking linearizability to the problem of checking that in this cross-product, the observer cannot reach a state where the semantics of the set data structure has been violated. To verify that that the observer cannot reach a state where a violation is reported, we compute a symbolic representation of an invariant that is satisfied by all reachable configuration of the cross-product of a program and an observer.

Future work
There are three main possible lines of future work we would like to work on from this thesis. In paper I, we conjecture that our method generalises to handle other types of properties in the data domain (e.g., comparing sets of stored values) or other types of constraints (e.g., constraints over lengths of lists or branches in a tree needed to express, e.g., balancedness of a tree). We
are currently working on an extension of FAs that can express more general classes of shapes (e.g., B+ trees) by allowing recursive nesting of boxes, and employing the CEGAR loop of ARTMC. We also plan to combine the method with techniques to handle concurrency. In paper II, we intend to extend the framework to more complex data structures such as trees and hash-maps, we also plan to define a mechanism that allow the automatic synthesis of both observers and controllers to make the framework fully automated. Finally, in paper III, we would like to is to extend the view abstraction to multi-threaded programs running on machines with different memory models. Such hardware systems employ store buffers and cache systems that could be modelled using views. This is an interesting challenge since it would help programmers to write their code under a given memory model that is simpler to reason around, and verify that the behaviour of the program is the same under another less-restricted memory model.
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