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#### A THESIS

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# Abstract

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# Introduction

1.1 Related Work

 $({\rm Under\ work...})$ 

1.2 Statement of Results

## Model of Computation and Definitions

In this chapter, we will describe our model of computation and give the definitions, which are based on Herlihy and Wing's [2] and Golab, Hadzilacos and Woelfel's [1].

The computational model we consider is the standard asynchronous shared memory model with a set of n processes, denoted as  $\{p_1, p_2, ..., p_n\}$ , where up to n-1 processes may fail by crashing.

**Type and Object.** A type  $\tau$  is defined as an automaton as follows [1],

$$\tau = (\mathcal{S}, s_{init}, \mathcal{O}, \mathcal{R}, \delta)$$

where  $\mathcal{S}$  is a set of states,  $s_{init} \in \mathcal{S}$  is the initial state,  $\mathcal{O}$  is a set of operation types,  $\mathcal{R}$  is the set of responses, and  $\delta : \mathcal{S} \times \mathcal{O} \to \mathcal{S} \times \mathcal{R}$  is a state transition mapping.

An object is an implementation of a type. For each type  $\tau$ , the transition mapping  $\delta$  captures the behaviour of objects of type  $\tau$ , in the absence of concurreny, as follows: if a process applies an operation of type opt to an object of type  $\tau$  which is in state s, the object may return to the process a response rsp and change its states to s' if and only if  $(s', rsp) \in \delta(s, opt)$ .

An object that supports only read() and write(x) operations is called a  $read/write\ register$  (or just register). Operation read() returns the current state of the register and leaves the state unchanged. Operation write(x) changes the state of the register to x.

An object that supports read() and CAS(x,y) operations is called *compare-and-swap* (CAS) object. Operation read() is the same as defined above. Operation CAS(x,y) changes the state of the object if and only if the current state is equal to x and then operation CAS(x,y) succeeds, and the state is changed to y and true is returned. Otherwise, operation CAS(x,y)

fails, the current state remains unchanged and false is returned.

**History** A history H, obtained by processes executing operations on objects, is a sequence of invocation and response events.

An invocation event is a 5-tuple,

$$INV = (invocation, p, obj, opt, t)$$

where *invocation* is the event type, p is the process executing the operation, obj is the object on which the operation is executed, opt is the operation type and t is the time when INV happens which is defined as the position of event INV in history H.

A response event is also a 5-tuple,

$$RSP = (response, p, obj, rsp, t)$$

where response is the event type, p is the process receiving response rsp from an oeration on object obj and t is the time when RSP happens which is defined as the position of event RSP in history H.

Response event  $(response, p_j, obj_q, rsp, t_1)$  matches invocation event  $(invocation, p_i, obj_p, opt, t_0)$  in history H, if the two events are applied by the same process to the same object, i.e, i = j and p = q. In this case, the response event is also called the matching response of the invocation event.

An operation execution in H is a pair oe = (INV, RSP) consisting of an invocation event INV and the next matching response event RSP, or just an invocation event INV, denoted as oe = (INV, null). In the latter case, we say the operation execution is pending. In the former case, we say the operation execution is complete. A history H is complete if all operation executions in H are complete, otherwise, it is incomplete.

History H' is an extension of history H if H is a prefix of H'. History H' is a completion of

history H if H' contains all the events in H and H' is an extension of H, and each operation execution in H' is complete.

H|obj of history H is the subsequence of all invocation and response events in H on object obj. If all invocation and response events in a history H have the same object name obj, then the H|obj = H.

Let H be a complete history. We associate a time interval  $I_{oe} = [t_0, t_1]$  with each operation execution oe = (INV, RSP) in H, where  $t_0$  and  $t_1$  are the points in time when INV and RSP happen. Similarly, for an incomplete history, we denote the time interval  $I_{oe}$  with respect to a pending operation execution oe = (INV, null) by  $I_{oe} = [t_0, \infty]$ .

Operation execution  $oe_0$  precedes operation execution  $oe_1$  in H if the response event of  $oe_0$  happens before the invocation event of  $oe_1$  in H. We say that  $oe_0$  and  $oe_1$  are concurrent in H if neither precedes the other.

A history is *sequential* if its first event is an invocation event, and each invocation event, except possibly the last one, is immediately followed by a matching response event.

**Linearization.** A history H linearizes to a sequential history S, if and only if S satisfies the following conditions: (1) S and any completion of H have the same operation executions, (2) sequential history S is valid, and (3) there is a mapping from each time interval  $I_{oe}$  to a time point  $t_{oe} \in I_{oe}$ , such that the sequential history S is obtained by sorting the operations in H based on their  $t_{oe}$  values.

A history is linearizable if and only if H linearizes to some sequential history S. In this case, S is called the linearization of H. For each operation opt in history H, we call time point  $t_{oe}$ , which is defined as above, the linearization point of opt. An object obj is linearizable if every history H on obj is linearizable.

Randomness. A process can execute local coin flip operations that return integer values

distributed uniformly at random from an arbitrary finite set of integers. In the following discussion, we use method random(s) to return a value which is distributed uniformly at random from set  $\{0, 1, 2, ..., s - 1\}$ .

Adversary. In the standard shared memory model, each process executes its program by applying shared memory operations (read(), write(x) CAS(x,y), etc) on objects, as determined by their program. Operation executions of concurrent processes can be interleaved arbitrarily.

For deterministic shared memory algorithms, we analysize their performance via the standard worst-case analysis assuming that this interleaving occurs in the "worst" possible way. However, in the analysis of randomized algorithms, we use an *adversary* model to describe how the interleaving of executions can be influenced by random choices made by processes.

There are several adversary models with different strengths The most pessimistic reasonable assumption is modeled by the *strong adaptive adversary*. We analyze our randomized shared memory algorithm under the assumption of a strong adaptive adversary, i.e, at any point of time, it can see the entire past history and knows the states of all processes and decides on the schedule accordingly.

## Dynamic Task Allocation Object

The dynamic task allocation (DTA) type supports two operations, DoTask() and InsertTask( $\ell$ ), where  $\ell$  is the identifier that is unique for each task (define what is task, give the domain).

Now we formalize the notion of type DTA by specifying the above two operations. We assume that there exists an atomic operation  $PutTask(M, \ell)$ , and a process associates (what is associate?) task  $\ell$  with memory location (what is memory location?) M by calling  $PutTask(M, \ell)$ . It returns success if task  $\ell$  is associated with location M, and returns failure if location M was already associated with another task.

Similarly, we assume there exists an atomic operation  $\operatorname{TryTask}(M)$ , and  $\operatorname{task} \ell$  associated with memory location M could be performed atomically by calling  $\operatorname{TryTask}(M)$ . Out of several processes calling  $\operatorname{TryTask}(M)$ , one receives  $\operatorname{success}$  and the index  $\ell$  of that  $\operatorname{task}$ , while all the others receive  $\operatorname{failure}$ . (state transition for all initial states, what happens next if there is  $\operatorname{no...}$ )

A task is *done* or *performed* if its index has been returned by a process after calling DoTask(). A task is *available* at location M if it has been inserted to M and *success* is returned by a process after calling  $InsertTask(\ell)$ , but is not done yet. A task is *available*, if it is *available* at some memory location.

The aim of operation DoTask() is to perform an available task on location M by calling TryTask(M). Every DoTask() may perform several TryTask(M) operations. However, only one of them will succeed. Once one TryTask(M) succeeds, then there is no available task on M and the task index  $\ell$  will be returned by DoTask(). Additionally, if there is no available task, then operation DoTask() returns  $\bot$ .

The goal of  $InsertTask(\ell)$  operation is to find a free memory location M and insert task  $\ell$  atomically by calling  $PutTask(M,\ell)$ .  $PutTask(M,\ell)$  fails if location M has been associated with another task, so each  $InsertTask(\ell)$  operation may perform several  $PutTask(M,\ell)$  operations, but only one of them will succeed. Once one TryTask(M) succeeds, then task  $\ell$  is available on location M and success notification is returned by  $InsertTask(\ell)$  operation. Type DTA is required to satisfy: (Validity) If a DoTask() operation returns  $\ell$ , then before the DoTask() operation, an  $InsertTask(\ell)$  was executed and returned success. (Uniqueness) Each task is performed at most once, i.e, for each task  $\ell$ , at most one DoTask() operation returns  $\ell$ .

In addition, the property that every inserted task is eventually done is also a desired progress property of the implementation of type DTA.

## Implementation of DTA

(under work...)

#### Method 1: DoTask()

```
1 while true do
        v \leftarrow root;
 2
        if v.surplus() \leq 0 then
 3
            return \perp;
 4
        end
 5
        /* Descent */;
 6
        while v is not a leaf do
 7
             (x_L, y_L) \leftarrow v.left.read();
 8
            (x_R, y_R) \leftarrow v.right.read();
 9
            s_L \leftarrow min(x_L - y_L, 2^{height(v)});
10
            s_R \leftarrow min(x_R - y_R, 2^{height(v)});
11
            r \leftarrow random(0, 1);
12
            if (s_L + s_R) = 0 then
13
                 Mark-up(v);
14
            else if r < s_L/(s_L + s_R) then
15
                v \leftarrow v.left;
16
17
            else
                v \leftarrow v.rght;
18
            end
19
        end
20
        /* v is a leaf */;
21
        (x,y) \leftarrow v.read();
22
        (flag, l) \leftarrow v.TryTask(task[y + 1]);
\mathbf{23}
        /* Update Insertion Count */;
\mathbf{24}
        v.CAS((x, y), (x, y + 1));
25
        v \leftarrow v.parent;
26
        Mark-up(v);
27
        if flag = success then
28
           \mathbf{return}\ \ell
29
        end
30
31 end
```

#### Method 2: InsertTask( $\ell$ )

```
32 while true do
        v \leftarrow root:
33
        /* Descent */;
34
        while v is not a leaf do
35
            (x_L, y_L) \leftarrow v.left.read();
36
            (x_R, y_R) \leftarrow v.right.read();
37
            s_L \leftarrow 2^{height(v)} - min(x_L - y_L, 2^{height(v)});
38
            s_R \leftarrow 2^{height(v)} - min(x_R - y_R, 2^{height(v)});
39
            r \leftarrow random(0, 1);
40
            if (s_L + s_R) = 0 then
41
                Mark-up(v);
42
            else if r < s_L/(s_L + s_R) then
43
                v \leftarrow v.left;
44
            else
45
                v \leftarrow v.rght;
46
            end
47
        end
48
        /* v is a leaf */;
49
        (x,y) \leftarrow v.read();
50
        flaq \leftarrow v. PutTask(task[x+1]);
51
        /* Update Insertion Count */;
52
        v.CAS((x, y), (x + 1, y));
53
        v \leftarrow v.parent;
54
        Mark-up(v);
55
        if flaq = success then
56
            return success
57
        end
58
59 end
```

#### Method 3: Mark-up(v)

```
60 if v is not null then

61 | for (i = 0; i < 2; i + +) do

62 | (x, y) \leftarrow v.read();

63 | (x_L, y_L) \leftarrow v.left.read();

64 | (x_R, y_R) \leftarrow v.right.read();

65 | v.CAS((x, y), (max(x, x_L + x_R), max(y, y_L + y_R));

66 | end

67 end
```

### Correctness Proof

#### 4.1 Correctness

The standard correctness condition for shared memory algorithms is linearizability, which was introduced by Herlihy and Wing in 1990 [2]. The intuition of linearizability is that real-time behavior of method calls must be preserved, i.e, if one method call precedes another, then the earlier call must have taken effect before the later one. By contrast, if two method calls overlap, we are free to order them in any convenient way since the order is ambiguous. Informally, a concurrent object is linearizable if each method call appears to take effect instantaneously at some moment between its invocation and response.

#### 4.1.1 Analysis and Proofs

By the definitions in Subsection 3.1.1, one way to show an object obj is linearizable is to prove every history H of obj is linearizable. Thus, we need to identify for each DoTask and InsertTask operation op (i.e, interval  $I_H(op)$ ) in H a linearization point  $t_H(op)$ , and prove that the sequential history S obtained by sorting these operations according to their  $t_H(op)$  satisfies the sequential specification  $S_{obj}$  of obj.

We notice that each complete DoTask or InsertTask operation can be associated with a unique task array slot based on the task it removed or inserted. Additionally, the removal and insertion count are both monotonically increasing. Thus, we could associate the node counts with operations which have been propagated to that node.

Now we define "an operation is counted at a node" recursively to formalize the operation

propagation.

A DoTask operation is counted at leaf v when the removal count of v is updated with the index of the task array slot where the performed task is located. Symmetrically, an InsertTask operation is counted at v when the insertion count of v is updated with the index of the task array slot where the inserted task is located.

Now we only define DoTask operation is counted at an inner node v because counting an InsertTask operation is symmetric as well.

Recall that the removal count of v is updated though CAS operation (line 6, method 3). Actually there could be more than one operations updating the count with the same value. We linearize all such CAS operations, which update the removal count of v with the same value y. We say for all these operations, only the first one in the linearization order counts the corresponding DoTask operation. In another word, a DoTask operation is counted at an inner node v as soon as the CAS updating operation that counts the DoTask is linearized. Based this definition, no operation will be counted twice at a node.

Please note that, the CAS operation counting the DoTask at node v is not necessary performed by the DoTask operation itself, i.e, suppose process p executes a DoTask operation and has successfully performed task  $\ell$  at certain leaf. Then the CAS operation counting this p.DoTask() at node v could be a different process q as long as q updates the removal count first in the linearization order.

Given the above concepts and properties, we could prove the following result: (under work...almost done)

- 4.2 Performance
- 4.2.1 DoTask Analysis
- 4.2.2 InsertTask Analysis

## Bibliography

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