Cryptography Meets Algorithms (15893) Lecture Notes

# Lecture 10: Hierarchical ORAM

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March 17, 2024

In an earlier lecture, we learned how to construct oblivious sorting, including bitonic sort which has $\mathcal{O}(n \log^2 n)$ cost [Bat68], bucket oblivious sort [ACN+] which has $\mathcal{O}(n \log n (\log \log n)^2)$ cost, and we showed that we can future eliminate the $(\log \log n)^2$ factor using techniques from [RS21, GWC+23].

We will cover Hierarchical ORAM [GO96] in today's lecture. Chronologically, Hierarchical ORAM was actually the original ORAM construction first proposed by Goldreich and Ostrovsky [GO96]. The version we will describe in today's class is *an optimized variant* of Goldreich and Ostrovsky's original construction [GO96] introduced in Chan et al. [CGLS17]. Since hierarchical ORAM consists of a hierarchy of oblivious hash tables, let's first introduce oblivious hash table.

## 1 Oblivious Hash Table

We consider a standard static hash table abstraction.

**Definition 1** (Static Hash Table)**.** Consider a data array $A = \{(k_i, v_i) | \perp\}_{i \in [N]}$, i.e., each element is either a *real* key-value pair denoted $(k_i, v_i)$ or a *filler* denoted $\perp$. It is guaranteed that all real elements have distinct keys. An oblivious hash table contains the following operations:

- $\mathsf{Build}(A)$: Given a data array $A$, output a data structure.
- $\mathsf{Lookup}(k)$: Given a key $k$, output $v$. If $k = \perp$ or $k \notin A$, then output $v = \perp$; otherwise the output $v$ satisfies $(k, v) \in A$.

The hash table we will construct satifies obliviousnes only if the same real key is never looked up twice. It turns out that such a hash table is sufficient for constructing hierarchical ORAM. We now formally define this non-recurrent requirement. A sequence of $\mathsf{Lookup}$ operations denoted $op$ is said to be *non-recurrent* if every real key requested is distinct. However, the sequence $op$ is allowed to make filler requests $k = \perp$ multiple times.

**Obliviousness.** Obliviousness requires the following: $\forall A_0, op_0, A_1, op_1$ s.t. $|A_0| = |A_1|, |op_0| = |op_1|$ and both $op_0$ and $op_1$ satisfy the non-recurrent constraint, then it holds that

$$\text{AccessPattern}(A_0, op_0) \approx \text{AccessPattern}(A_1, op_1)$$

where $\text{AccessPattern}(A, op)$ denotes the access patterns emitted by the oblivious hash table when making a single call to $\mathsf{Build}(A)$ followed by a sequence of $\mathsf{Lookup}$ operations denoted $op$, and $\approx$ means computational indistinguishability. More intuitively, we want that for any two data arrays and operation sequences of the same length, the access patterns in oblivious hash tables are indistinguishable as long as the same key is never looked up twice.

**Construction.** Now let's introduce how to build an oblivious hash table. We use the balls and bins hashing with $N$ elements (balls) and $N/Z$ bins[1], where $Z = \omega(\log N)$. The expected number of elements in any fixed bin is $Z$. With Chernoff bound, the probability of any bin having more than $2Z$ elements is negligible in $N$. Thus, we set the bin size to be $2Z$. One way to randomly assign the element $(k, v)$ to a bin is to use a persudorandom function $PRF_{sk}(k)$, where the $sk$ is the secret key held by the ORAM client and the PRF output range is $[N/Z]$, indicating the desired bin index.

So now, we want to *obliviously* throw the balls into the bins without revealing which bin each ball goes to. At the end, each bin is padded with filler elements to the maximum capacity. To achieve this, we will use the "oblivious random bin assignment" construction from the last lecture — recall from the last lecture that oblivious random bin assignment was a key building block for constructing an oblivious random permutation, which in turn served as a key building block in the construction of bucket oblivious sort. More specifically, the initial labels of an element with key $k$ is computed by $PRF_{sk}(k)$, and then we use a butterfly network (where all intermediate bins are of size $2Z$) to route all elements to their desired destination bins.

To implement $\mathsf{Lookup}(k)$, if $k$ is a real key, the client simply reads the entire bin indexded $PRF_{sk}(k)$. If $k = \perp$, then the client reads a random bin.

**Analysis.** Building an oblivious hash table calls the "oblivious random bin assignment" algorithm of the last lecture. Recall that in the last lecture, we mentioned that using techniques from [RS21, GWC+23], we can accomplish this with $\mathcal{O}(N \log N)$ cost. For every lookup, the cost is the bin size, which is $Z = \omega(\log N)$. Different from (non-oblivious) static hashing that has constant cost for each lookup, our oblivious (static) hash table has a bit higher overhead.

## 2 Hierarchical ORAM

So now with this oblivious hash table, we can construct a Hierarchical ORAM. The hierachical ORAM essentially uses Bentley and Saks' hierarchical data structure idea [BS80] that converts a static data structure (in our case, an oblivious hash table) into a dynamic one.

In Hierarchical ORAM, we have a hierarchy of exponentially growing oblivious hash tables as shown in Figure 1. Each level of the hierarchy has a different capacity, and the capacity grows exponentially with the level. The levels are indexed $0, 1, \ldots, L$, and the $i$-th level's capacity is $2^i$. Additionally, every level has a label, either **empty** or **full**. "Empty" means that this level current is not being used, and "full" means that this level is currently in use. Whether each level is empty or full only depends on the time step (i.e., how many operations have been performed so far), so this information is public and is known by the adversary.

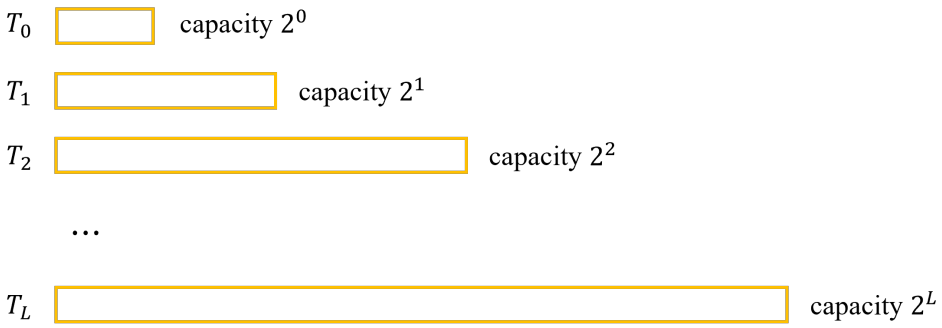


Figure 1: Hierarchical ORAM

[1]For simplicity, we assume that $N$ is divisible by $Z$.

**Algorithm 1:** HIERARCHICAL ORAM

**Input:** `addr`, an address to lookup in the hash tables; `op`, the operation `write` or `read`; `data`, the data to write.

**Output:** $data^*$, the lookup result, if `op` = `read`.

```
found ← false;
for l ← 0 to L do
    if not found then
        fetched ← T_l.Lookup(addr);
        if fetched ≠ ⊥ then
            found ← true;
            data* ← fetched;
    else
        T_l.lookup(⊥)
if op = read then
    Let T^φ ← {(addr, data*)};
else if op = write then
    Let T^φ ← {(addr, data)};
Let l be the smallest level such that T_l is empty, if all levels are full, let l = L;
Let S ← T^φ ∪ T_0 ∪ ··· ∪ T_{l−1};
If all levels are full, let S ← S ∪ T_L;
T_l.Build(SuppressDup(S));
Mark T_l as full, mark T_0, ··· , T_{l−1} as empty;
if op = read then
    return data*;
```

Recall that we want to implement a memory abstraction with the address space $1 \dots N$. Initially, we may assume that all memory cells are empty and contain the symbol $\perp$. Therefore, initially all the ORAM levels are marked empty. However, later, as we perform read and write operations, at any point of time, some levels might be full and others are empty.

Now suppose a sequence of operations arrive. There are two types of operations, `Read(addr)` and `Write(addr, data)`.

**Intuition.** Before diving into the detailed algorithm, let's first introduce the high-level idea of how to perform these operations.

If we want to look up some element, we will start from the smallest level. Since every level is either empty or full. If it's full, look up that hash table for the desired element. There are two outcomes: either we find the element, or we don't find the element. If we have found the element, for all future levels, we just perform a fake lookup by requesting a filler element $\perp$. Similar to the tree-based ORAM, we will also need to relocate the retrieved element. So in Hierarchical ORAM, we will put the retrieved element to the smallest level if this level is empty. If this level is not empty, we will need to merge the retrieved element with the existing element in this level and and merged hash table of size 2 is copied to the next level if it is empty. However, if the next level is not empty, then it causes a cascading merge until we find a level that is empty or we reach the largest level.

**Algorithm.** The detailed algorithm is shown in Algorithm 1. We use $T_0, \dots T_L$ to denote the oblivious hash table of the $L + 1$ levels. The algorithm starts by initiating a binary variable `found` to `false`. Then, for each level, we check if the element is in the hash table. If we find the

element, we set `found` to `true` and store the element in $\texttt{data}^*$. If the element is already found, we just look up a filler for all future levels — importantly, *this guarantees that the non-recurrent constraint is satisfied for every oblivious hash table.* After we go through all levels, the fetched element is in $\texttt{data}^*$.

If the operation `op` is `read`, we will need to write back the retrieved element $T^\phi = \{(\texttt{addr}, \texttt{data}^*)\}$. Otherwise, if the operation `op` is `write`, we will need to write back the new element $T^\phi = \{(\texttt{addr}, \texttt{data})\}$. Let $l$ be the smallest level such that $T_l$ is empty, if all levels are full, let $l = L$. We want to merge all levels $0 \ldots l-1$ along with the new element into level $l$ (or merge all levels $0 \ldots L$ along with the new element into the largest level $L$). Then we obliviously build the hash table at level $l$ with the input set $S$. After the build step, we mark the level $l$ as full and the levels 0 to $l-1$ as empty.

Note that if the oblivious hash table doesn't remove the element we just looked up, there might be duplicate elements in the set $S$ and we will need to suppress the duplicate elements. The elements in the smaller level is always fresher than the elements in the larger level, so we can just keep the fresher element and remove the older one. We can use the oblivious sorting to do the duplicate suppression, denoted $\mathsf{SuppressDup}$ in Algorithm 1. Specifically, we obliviously sort the elements by key and by freshness and removes the duplicate elements using a linear scan.

**Analysis.** Now let's analyze the overhead of Hierarchical ORAM. For each request, there are two sources of the overhead: the lookup phase, and the rebuild phase. The lookup cost is $L \times Z$, the number of levels $L$ times the lookup cost of at each level, which is at most the bin size $Z$. Recall we can set $Z = \log N \cdot \alpha(N)$, where $\alpha(N)$ is an arbitrarily small super-constant function. Thus the lookup cost is $\mathcal{O}(\log^2 N \cdot \alpha(N))$.

For rebuild cost, a level of size $2^i$ is rebuilt every $2^i$ steps. So to rebuild a level of size $2^i$, the cost is $2^i \cdot \log 2^i$ (see the analysis of the oblivious hash table). By amortizing the cost to each step, every level has a cost $\leq \log N$. Thus, the rebuild cost is $\mathcal{O}(\log^2 N)$.

Summarizing the above, each memory request has cost $\mathcal{O}(\log^2 N \cdot \alpha(N))$ where $\alpha(N)$ may be an arbitrarily small super-constant function. Note that the original Goldreich-Ostrovsky construction [GO96] has a cost of $\mathcal{O}(\log^3 N)$, because that they throw $N$ balls into $N$ bins, instead of $N/Z$ bins.

## References


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