## Advanced Cryptography CS 655

## Week 15:

- Quantum Random Oracle Model
- Oblivious RAM


## Homework 4 Released

Course Presentations: Next Week (Schedule Announced Soon)

## Recap

## - Quantum Basics

- Grover Search
- Quantum Random Oracle Model
- Useful Results
- $\psi=\sum_{x, y} \alpha_{x, y}|x, y\rangle$ and $\psi^{\prime}=\sum_{x, y} \beta_{x, y}|x, y\rangle \rightarrow$ Measurement can
distinguish two states with probability at most $4 \sqrt{\sum_{x, y}\left|\alpha_{x, y}-\beta_{x, y}\right|^{2}}$
- Upper bound Euclidean distance between final states $\psi$ and $\psi^{\prime}$ when we use oracles H and $\mathrm{H}^{\prime}$ in terms of "query magnitude" on bad inputs x where $H(x) \neq H^{\prime}(x)$.


## State in QROM

- Typically we write

Output Register

$$
\begin{gathered}
|\Psi\rangle=\sum_{\substack{x, y, z \\
\text { Input Register }}} \alpha_{x, y, z}|x, y, z\rangle \\
\mathrm{RO}|\Psi\rangle=\sum_{x, y, z} \alpha_{x, y, z}|x, R O(X) \oplus y, z\rangle
\end{gathered}
$$

## Query Magnitude

- Let $S \subset\{0,1\}^{n}$ be a set of inputs and let $\psi=\sum_{x, y} \alpha_{x, y}|x, y\rangle$ be a quantum state. Then

$$
Q M(\psi, S):=\sum_{x \in S} \sum_{y, z} \alpha_{x, y, z}^{2}
$$

- If $A^{H(.)}(w)$ generates states $\psi_{0, H, w}, \psi_{1, H, w}, \ldots, \psi_{T, H, w}$ we can write

$$
Q M(A, H, w, S):=\sum_{i<T} Q M\left(\Psi_{i, H, w}, S\right)
$$

Theorem: If $H(x)=H^{\prime}(x)$ for all inputs $x \notin S$ then the Euclidean distance between the final states $\psi_{T, H, w}$ and $\psi_{T, H \prime, w}$ is at most T • $Q M(A, H, w, S)$

## Homework Hint

- Intuition: for a random/small sets S we expect $Q M(A, H, w, S)$ to be small with high probabilty.
- If $S_{1}, \ldots S_{r} \subset\{0,1\}^{n}$ are disjoint sets of inputs then

$$
\sum_{i \leq r} Q M\left(\psi, S_{i}\right)=\sum_{i \leq r} \sum_{x \in S} \sum_{y, z} \alpha_{x, y, z}^{2} \leq \sum_{x} \sum_{y, z} \alpha_{x, y, z}^{2}=1
$$

- If $A^{H(.)}(w)$ generates states $\psi_{0, H, w}, \Psi_{1, H, w}, \ldots, \Psi_{T, H, w}$ we can write

$$
\begin{gathered}
Q M(A, H, w, S):=\sum_{i<T} Q M\left(\Psi_{i, H, w}, S\right) \\
\sum_{i \leq r} Q M\left(A, H, w, S_{i}\right) \leq T
\end{gathered}
$$

## Quantum Computing: Useful Theorem

Let $\psi=\sum_{x, y} \alpha_{x, y}|x, y\rangle$ and $\psi^{\prime}=\sum_{x, y} \beta_{x, y}|x, y\rangle$

Fix a quantum measurement and let $\mathfrak{D}$ and $\mathfrak{D}^{\prime}$ be the distribution of outputs when we measure quantum states $\psi$ and $\psi^{\prime}$ respectively.

Definition: Total Variation Distance

$$
\operatorname{TVD}\left(\mathfrak{D}, \mathfrak{D}^{\prime}\right):=\sum_{w}\left|\operatorname{Pr}_{\mathfrak{D}}[w]-\operatorname{Pr}_{\mathfrak{D}^{\prime}}[w]\right|
$$

## Quantum Computing: Useful Theorem

Theorem: Let $\psi=\sum_{x, y} \alpha_{x, y}|x, y\rangle$ and $\psi^{\prime}=\sum_{x, y} \beta_{x, y}|x, y\rangle$ be two quantum states. Fix any measurement and let $\mathfrak{D}$ and $\mathfrak{D}^{\prime}$ be the distribution of outputs when we measure quantum states $\psi$ and $\psi^{\prime}$ respectively. We have

$$
\operatorname{TVD}\left(\mathfrak{D}, \mathfrak{D}^{\prime}\right):=\sum_{w}\left|\operatorname{Pr}_{\mathfrak{D}}[w]-\mathfrak{D}_{[w]}^{\prime}[w]\right| \leq 4 \sqrt{\sum_{x, y}\left|\alpha_{x, y}-\beta_{x, y}\right|^{2}}
$$

Intuition: WHP we cannot distinguish between "close" states $\psi$ and $\psi^{\prime}$ with any measurement

## Phase Oracle vs Standard Oracle

- Typically we write

Output Register

$$
|\Psi\rangle=\sum_{x, y, z} \alpha_{x, y, z}|x, y, z\rangle
$$

Input Register
Auxilliary State (not associated with current RO query)

$$
\mathrm{StO}|\Psi\rangle=\sum_{x, y, z} \alpha_{x, y, z}|x, R O(x) \oplus y, z\rangle
$$

Phase Oracle

$$
\operatorname{PhO}|\Psi\rangle=\sum_{x, y, z} \alpha_{x, y, z}(-1)^{y \cdot R O(x)}|x, y, z\rangle
$$

## Equivalence: Phase/Standard Oracle

Lemma 3. For any adversary $A$ making queries to StO , let $B$ be the adversary that is identical to $A$, except it performs the Hadamard transformation $\mathrm{H}^{\otimes n}$ to the response registers before and after each query. Then $\operatorname{Pr}\left[A^{\mathrm{StO}}()=1\right]=$ $\operatorname{Pr}\left[B^{\mathrm{PhO}}()=1\right]$

$$
\begin{aligned}
\text { StO| }|\psi\rangle & =\sum_{x, y, z} \alpha_{x, y, z}|x, R O(x) \oplus y, z\rangle \\
\text { Pho }|\psi\rangle & =\sum_{x, y, z} \alpha_{x, y, z}(-1)^{y \cdot R o(x)}|x, y, z\rangle
\end{aligned}
$$

Let $\operatorname{Had}_{\text {out }}=I^{\otimes n} H^{\otimes n} I^{\otimes|z|}$ be a unitary transformation which applies the Hadamard transform to response registers ( y ) and identity transform elsewhere. Then

$$
\operatorname{Had}_{\mathrm{out}} \operatorname{PhO}\left(\operatorname{Had}_{\mathrm{out}}|\Psi\rangle\right)=\mathrm{StO}|\psi\rangle=\sum_{x, y, z} \alpha_{x, y, z}|x, R O(x) \oplus y, z\rangle
$$

## Views of the Quantum Random Oracle

We can view a function $\boldsymbol{H}:\{\mathbf{0}, \mathbf{1}\}^{2 \lambda} \rightarrow\{\mathbf{0}, \mathbf{1}\}^{\lambda}$ as a string of length $\lambda \mathbf{2}^{2 \lambda}$
$H(x)$ simply returns the $\lambda$ bit string at locations $\mathbf{x} \lambda, \ldots,(x+\mathbf{1}) \lambda-\mathbf{1}$
Now we can view the state as

$$
|\psi\rangle \otimes|H\rangle=\sum_{x, y, z} \alpha_{x, y, z}|x, y, z\rangle \otimes|H\rangle
$$

Standard oracle performs map

$$
|x, y, z\rangle \otimes|H\rangle \rightarrow|x, y \oplus H(x), z\rangle \otimes|H\rangle
$$

Algorithm can only apply unitary transforms to first part of state $|x, y, z\rangle$

## Views of the Quantum Random Oracle

We can view a function $\boldsymbol{H}:\{\mathbf{0}, \mathbf{1}\}^{2 \lambda} \rightarrow\{\mathbf{0}, \mathbf{1}\}^{\lambda}$ as a string of length $\lambda 2^{2 \lambda}$

World 1 (StO'): Pick random function H and run algorithm with initial state $\left|\psi_{0}\right\rangle \otimes|H\rangle$

World 2 (StO): run algorithm with initial state (uniform superposition of all oracles)

$$
\left|\psi_{0}\right\rangle \bigotimes\left(\frac{1}{\sqrt{2^{\lambda 2^{2 \lambda}}}} \sum_{H}|H\rangle\right)
$$

## Views of the Quantum Random Oracle

World 1 (StO'): Pick random function H and run algorithm with initial state $\left|\psi_{0}\right\rangle \otimes|H\rangle$

World 2 (StO): run algorithm with initial state (uniform superposition of all oracles)

$$
\left|\psi_{0}\right\rangle \bigotimes\left(\frac{1}{\sqrt{2^{\lambda 2^{2 \lambda}}}} \sum_{H}|H\rangle\right)
$$

Lemma 2. StO and $\mathrm{StO}^{\prime}$ are perfectly indistinguishable. That is, for any adversary $A$ making oracle queries, let $A^{\mathrm{StO}}()$ and $A^{\mathrm{StO}^{\prime}}()$ denote the algorithm interfacing with StO and $\mathrm{StO}^{\prime}$, respectively. Then $\operatorname{Pr}\left[A^{\mathrm{StO}}()=1\right]=\operatorname{Pr}\left[A^{\mathrm{SOO}^{\prime}}()=1\right]$

## Another View

World $2^{\prime}:|H\rangle=\left\{(x, \perp): x \in\{0,1\}^{2 \lambda}\right\}$ where $\perp$ indicates that $H(x)$ is not yet assigned
Run algorithm with initial state $\left|\psi_{0}\right\rangle \otimes|H\rangle$
Oracle Map (Intuitions): if $H(x)=\perp$

$$
|x, y, z\rangle \otimes|H\rangle \rightarrow 2^{\lambda / 2} \sum_{w}|x, y \oplus w, z\rangle \otimes|H \cup(x, w)\rangle
$$

Where $H \cup(x, w)$ replaces $(x, \perp)$ with $(x, w)$
Idea: measuring red state yields a list of query/output pairs!

## Not Quite that Simple...

World $2^{\prime}:|H\rangle=\left\{(x, \perp): x \in\{0,1\}^{2 \lambda}\right\}$ where $\perp$ indicates that $H(x)$ is not yet assigned
Run algorithm with initial state $\left|\Psi_{0}\right\rangle \otimes|H\rangle$
Oracle Map (Intuitions): if $H(x)=\perp$

$$
|x, y, z\rangle \otimes|H\rangle \rightarrow 2^{\lambda / 2} \sum_{w}|x, y \oplus w, z\rangle \otimes|H \cup(x, w)\rangle
$$

Where $H \cup(x, w)$ replaces $(x, \perp)$ with $(x, w)$
Idea: measuring red state yields a list of query/output pairs!
Question: What if $H(x) \neq \perp$ ? How do we make sure oracle is unitary transformation?

## Towards Unitary Transform

World $2^{\prime}:|H\rangle=\left\{(x, \perp): x \in\{0,1\}^{2 \lambda}\right\}$ where $\perp$ indicates that $H(x)$ is not yet assigned
Run algorithm with initial state $\left|\Psi_{0}\right\rangle \otimes|H\rangle$

Build oracle out of unitary transforms: StdDecomp and StO'
StdOracle := StdDecomp o St0' $\circ$ StdDecomp
PhsOracle $:=$ StdDecomp $\circ \mathrm{PhsO}^{\prime} \circ$ StdDecomp
Where

$$
\begin{aligned}
\mathrm{StO}^{\prime}|x, y, z\rangle \otimes|H\rangle & \rightarrow|x, y \oplus H(x), z\rangle \otimes|H\rangle \\
\mathrm{PhsO}^{\prime}|x, y, z\rangle \otimes|H\rangle & \rightarrow(-1)^{y \cdot H(x)}|x, y, z\rangle \otimes|H\rangle
\end{aligned}
$$

Note: Define $y \oplus \perp:=y$ and $y \cdot \perp:=0$ for completeness

## Towards Unitary Transform

Build oracle out of unitary transforms: StdDecomp and StO'
StdDecomp $|x, y, z\rangle \otimes|H\rangle \rightarrow|x, y, z\rangle \otimes \operatorname{StdDecomp}_{x}|H\rangle$

$$
\operatorname{StdDecomp}_{x}|H\rangle=2^{-\frac{\lambda}{2}} \sum_{w}|H \cup(x, w)\rangle \text { if } H(x)=\perp
$$

Reversibility: If $H^{\prime}(x)=\perp$ then

$$
\operatorname{StdDecomp}_{x}\left(2^{-\frac{\lambda}{2}} \sum_{w}\left|H^{\prime} \cup(x, w)\right\rangle\right)=\left|H^{\prime}\right\rangle
$$

## Towards Unitary Transform

Build oracle out of unitary transforms: StdDecomp and StO'

$$
\operatorname{StdDecomp}|x, y, z\rangle \otimes|H\rangle \rightarrow|x, y, z\rangle \otimes \operatorname{StdDecomp}_{x}|H\rangle
$$

$$
\operatorname{StdDecomp}_{x}|H\rangle=2^{-\frac{\lambda}{2}} \sum_{w}|H \cup(x, w)\rangle \text { if } H(x)=\perp
$$

If $H^{\prime}(x)=\perp$ and $z \neq 0$ then

$$
\operatorname{StdDecomp}_{x}\left(2^{-\frac{\lambda}{2}} \sum_{w}(-1)^{z \cdot w}\left|H^{\prime} \cup(x, w)\right\rangle\right)=2^{-\frac{\lambda}{2}} \sum_{w}(-1)^{z \cdot w}\left|H^{\prime} \cup(x, w)\right\rangle
$$

## Reversibility:

$$
\operatorname{StdDecomp}_{x}\left(2^{-\frac{\lambda}{2}} \sum_{w}\left|H^{\prime} \cup(x, w)\right\rangle\right)=\left|H^{\prime}\right\rangle
$$

All pairs of the form ( $\mathrm{x}, \mathrm{w}$ ) are removed

## Compressed Oracles

World 2': $|H\rangle=\left\{(x, \perp): x \in\{0,1\}^{\lambda}\right\}$ where $\perp$ indicates that $H(x)$ is not yet assigned
Run algorithm with initial state $\left|\psi_{0}\right\rangle \otimes|H\rangle$

Build oracle out of unitary transforms: StdDecomp and StO' StdOracle := StdDecomp $\circ$ StO' $\circ$ StdDecomp
PhsOracle $:=$ StdDecomp $\circ \mathrm{PhsO}^{\prime} \circ$ StdDecomp

Compressed Versions: CPhsO and CPhsO
Idea: If we know there are only T queries then $|H\rangle$ will never have more than T entries that are not $\perp \rightarrow$ can compress representation of $|H\rangle$

## Compressed Oracles

## Compressed Versions: CPhsO and CPhsO

Idea: If we know there are only $T$ queries then $|H\rangle$ will never have more than T entries that are not $\perp \boldsymbol{\rightarrow}$ can compress representation of $|H\rangle$

```
Lemma 4. CStO and StO are perfectly indistinguishable. CPhsO and PhO are
perfectly indistinguishable. That is, for any adversary \(A\), we have \(\operatorname{Pr}\left[A^{\mathrm{CstO}}()=\right.\)
\(1]=\operatorname{Pr}\left[A^{\text {StO }}()=1\right]\), and for any adversary \(B\), we have \(\operatorname{Pr}\left[B^{\mathrm{CPhsO}}()=1\right]=\)
\(\operatorname{Pr}\left[A^{\mathrm{PhO}}()=1\right]\).
```


## A Helpful Lemma

Lemma 5. Consider a quantum algorithm $A$ making queries to a pandom oracle $H$ and outputting tuples $\left(x_{1}, \ldots, x_{k}, y_{1}, \ldots, y_{k}, z\right)$. Let $R$ be a collection of such tuples. Suppose with probability $p$, A outputs a tuple such that (1) the tuple is in $R$ and (2) $H\left(x_{i}\right)=y_{i}$ for all $i$. Now consider running $A$ with the oracle CStO , and suppose the database $D$ is measured after $A$ produces its output. Let $p^{\prime}$ be the probability that (1) the tuple is in $R$, and (2) $D\left(x_{i}\right)=y_{i}$ for all $i$ (and in particular $\left.D\left(x_{i}\right) \neq \perp\right)$. Then $\sqrt{p} \leq \sqrt{p^{\prime}}+\sqrt{k / 2^{n}}$

## Example:

$$
R_{\text {collision }}=\left\{\left(x_{1}, y\right),\left(x_{2}, y\right): x_{1}, x_{2} \in\{0,1\}^{2 \lambda} \wedge y \in\{0,1\}^{\lambda}\right\}
$$

$p$ denotes probability that A outputs a hash collision (regular QROM). $p^{\prime}$ denotes probability that we measure a database with some colliding pair (when using compressed oracle CStO)

## A Helpful Lemma

Lemma 5. Consider a quantum algorithm $A$ making queries to a random oracle $H$ and outputting tuples $\left(x_{1}, \ldots, x_{k}, y_{1}, \ldots, y_{k}, z\right)$. Let $R$ be a collection of such tuples. Suppose with probability $p$, A outputs a tuple such that (1) the tuple is in $R$ and (2) $H\left(x_{i}\right)=y_{i}$ for all $i$. Now consider running $A$ with the oracle CStO , and suppose the database $D$ is measured after $A$ produces its output. Let $p^{\prime}$ be the probability that (1) the tuple is in $R$, and (2) $D\left(x_{i}\right)=y_{i}$ for all $i$ (and in particular $\left.D\left(x_{i}\right) \neq \perp\right)$. Then $\sqrt{p} \leq \sqrt{p^{\prime}}+\sqrt{k / 2^{n}}$

## Example:

$$
R_{\text {collision }}=\left\{\left(x_{1}, y\right),\left(x_{2}, y\right): x_{1}, x_{2} \in\{0,1\}^{2 \lambda} \wedge y \in\{0,1\}^{\lambda}\right\}
$$

$p$ denotes probability that A outputs a hash collision (regular QROM).
$p^{\prime}$ denotes probability that we measure a database with some colliding pair Typically, much easier to upper bound $p^{\prime} \rightarrow$ upper bound for $p$ (quantity we want to upper bound)

## Grover Search Revisited

Theorem 1. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the probability it contains a pair of the form $\left(x, 0^{n}\right)$ is at most $O\left(q^{2} / 2^{n}\right)$.

Corollary 1. After making q quantum queries to a random oracle, the probability of finding a pre-image of $0^{n}$ is at most $O\left(q^{2} / 2^{n}\right)$.

Proof: Define $R_{0}=\left\{(x, 0): x \in\{0,1\}^{2 \lambda}\right\} \rightarrow$ Let $p$ denote probability outputs $(\mathrm{x}, \mathrm{y}=0) \in R_{0}$ with $\mathrm{H}(\mathrm{x})=\mathrm{y}=0$ i.e., found a pre-image
Theorem 1 shows that $p^{\prime}=O\left(\frac{q^{2}}{2^{\lambda}}\right)$. Now applying Lemma 5 we have

$$
\sqrt{p} \leq O\left(\sqrt{\frac{q^{2}}{2^{\lambda}}}\right)+2^{-\frac{\lambda}{2}}=O\left(\frac{q}{\sqrt{2^{\lambda}}}\right) \Rightarrow p=O\left(\frac{q^{2}}{2^{\lambda}}\right)
$$

## Hash Collisions

Theorem 2. For any adversary making $q$ queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Corollary 2. After making q quantum queries to a random oracle, the probability of finding a collision is at most $O\left(q^{3} / 2^{n}\right)$.

Proof: $R_{\text {collision }}=\left\{\left(x_{1}, y\right),\left(x_{2}, y\right): x_{1}, x_{2} \in\{0,1\}^{2 \lambda} \wedge y \in\{0,1\}^{\lambda}\right\} \rightarrow$ Let $p$ denote probability attacker outputs $\left(x_{1}, y\right),\left(x_{2}, y\right) \in R_{\text {collision }}$ with $\mathrm{H}\left(\mathrm{x}_{1}\right)=\mathrm{H}\left(\mathrm{x}_{2}\right)=\mathrm{y}$ i.e., found a collision
Theorem 1 shows that $p^{\prime}=O\left(\frac{q^{3}}{2^{\lambda}}\right)$. Now applying Lemma 5 we have

$$
\sqrt{p} \leq O\left(\sqrt{\frac{q^{3}}{2^{\lambda}}}\right)+O\left(\frac{1}{\sqrt{2^{\lambda}}}\right)=O\left(\frac{q \sqrt{q}}{\sqrt{2^{\lambda}}}\right) \Rightarrow p=O\left(\frac{q^{3}}{2^{\lambda}}\right)
$$

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+S=I$
$P$ projects onto random oracles that contain a collision
Consider the state

$$
|\psi\rangle=\sum_{x, y, z, H} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle \Rightarrow P|\psi\rangle:=\sum_{H \in \text { Collision }} \sum_{x, y, z} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle
$$

Where

$$
\text { Collision }=\left\{H: \exists x_{1}, x_{2} \text { s.t. } H\left(x_{1}\right)=H\left(x_{2}\right) \neq \perp\right\}
$$

## Proof of Theorem 2

Theorem 2. For any adversary making $q$ queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+S=I$ $Q$ projects onto states that do contain a collision such that $y \neq 0$ and $H(x)=\perp$ Consider the state

$$
|\psi\rangle=\sum_{x, y, z, H} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle \Rightarrow Q|\psi\rangle:=\sum_{\substack{x, y \neq 0, z}} \sum_{\substack{H \notin \text { Collision, } \\ H(x)=\perp}} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle
$$

Where

$$
\text { Collision }=\left\{H: \exists x_{1}, x_{2} \text { s.t. } H\left(x_{1}\right)=H\left(x_{2}\right) \neq \perp\right\}
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Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+S=I$
$R$ projects onto states that do contain a collision such that $y \neq 0$ and $H(x) \neq \perp$ Consider the state

$$
|\psi\rangle=\sum_{x, y, z, H} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle \Rightarrow R|\psi\rangle:=\sum_{\substack{x, y \neq 0, z}} \sum_{\substack{H \notin \text { Collision, } \\ H(x) \neq \perp}} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle
$$

Where

$$
\text { Collision }=\left\{H: \exists x_{1}, x_{2} \text { s.t. } H\left(x_{1}\right)=H\left(x_{2}\right) \neq \perp\right\}
$$

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+S=I$
$S$ projects onto states that do contain a collision such that $y=0$
Consider the state

$$
|\psi\rangle=\sum_{x, y, z, H} \alpha_{x, y, y, H}|x, y, z\rangle \otimes|H\rangle \Rightarrow R|\psi\rangle:=\sum_{x, y=0, z} \sum_{H \notin \text { collision, }} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle
$$

Where

$$
\text { Collision }=\left\{H: \exists x_{1}, x_{2} \text { s.t. } H\left(x_{1}\right)=H\left(x_{2}\right) \neq \perp\right\}
$$

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+$ $S=I$
$S$ projects onto states that do contain a collision such that $y=0$ We want to bound $\| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle \|$ (Euclidean norm of "bad/collision state") after each random oracle query.

$$
\operatorname{CPhsO}|\psi\rangle=\operatorname{CPhsO}(P|\psi\rangle+Q|\psi\rangle+R|\psi\rangle+S|\psi\rangle)
$$

## Proof of Theorem 2

Theorem 2. For any adversary making $q$ queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

Proof: We first define projections $P, Q, R, S$ such that $P+Q+R+S=I$
$S$ projects onto states that do contain a collision such that $y=0$

$$
\begin{aligned}
& \| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle\|\leq\| \mathrm{P} \circ \mathrm{CPhsO} \circ P \circ|\psi\rangle \| \\
&+\| \mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle \| \\
&+\| \mathrm{R} \circ \mathrm{CPhsO} \circ R \circ|\psi\rangle \| \\
&+\| \mathrm{P} \circ \mathrm{CPhsO} \circ S \circ|\psi\rangle \|
\end{aligned}
$$

## Proof of Theorem 2

Theorem 2. For any adversary making $q$ queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

## Proof:

$\| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle\|\leq\| \mathrm{P} \circ \mathrm{CPhsO} \circ \mathrm{P} \circ|\psi\rangle \|$ $+\| \mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle \|$ $+\| \mathrm{R} \circ \mathrm{CPhsO} \circ \mathrm{R} \circ|\psi\rangle \|$
$+\| \mathrm{P} \circ \mathrm{CPhsO} \circ S \circ|\psi\rangle \| \quad$ Old Projection before Ro query

Fact 1: $\| \mathrm{P} \circ \mathrm{CPhsO} \circ P \circ|\psi\rangle\|\leq\| \mathrm{CPhsO} \circ P \circ|\psi\rangle\|=\| P \circ|\psi\rangle \|$

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

## Proof:

$$
\begin{aligned}
\| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle \| \leq & \| P \circ|\psi\rangle\|+\| \mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle \| \\
& +\| \mathrm{P} \circ \mathrm{CPhsO} \circ R \circ|\psi\rangle \| \\
& +\| \mathrm{P} \circ \mathrm{CPhsO} \circ S \circ|\psi\rangle \|
\end{aligned}
$$

Fact 2: $\| \mathrm{P} \circ(\mathrm{CPhsO} \circ S \circ|\psi\rangle)\|\leq\| \mathrm{P} \circ S \circ|\psi\rangle \|=0$
CPhsO does not modify states in projection $S \circ|\psi\rangle$ ! States in projection S do no have collision!
$\rightarrow \mathrm{CPhsO} \circ S \circ|\psi\rangle=S \circ|\psi\rangle$

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

## Proof:

$$
\begin{aligned}
\| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle \| \leq & \| P \circ|\psi\rangle\|+\| \mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle \| \\
& +\| \mathrm{P} \circ \mathrm{CPhsO} \circ R \circ|\psi\rangle \|+0
\end{aligned}
$$

Fact 3: $\| \mathrm{P} \circ(\mathrm{CPhsO} \circ R \circ|\psi\rangle)\left\|\leq \sqrt{\frac{q}{2^{\lambda}}}\right\| R \circ|\psi\rangle \|$
Proof: Skipped (similar to Fact 4)

## Proof of Theorem 2

Theorem 2. For any adversary making q queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

## Proof:

$$
\| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle\|\leq\| P \circ|\psi\rangle\left\|+\sqrt{\frac{q}{2^{\lambda}}}\right\| R \circ|\psi\rangle\|+\| \mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle \|
$$

Fact 4: $\| \mathrm{P} \circ(\mathrm{CPhsO} \circ Q \circ|\psi\rangle)\left\|\leq \sqrt{\frac{q}{2^{\lambda}}}\right\| Q \circ|\psi\rangle \|$

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$$
\begin{aligned}
& \| \mathrm{P} \circ \mathrm{CPhsO} \circ|\psi\rangle\|\leq\| P \circ|\psi\rangle\left\|+\sqrt{\frac{q}{2^{2}}}\right\| R \circ|\psi\rangle\left\|+\sqrt{\frac{q}{2^{\lambda}}}\right\| Q \circ|\psi\rangle \| \\
& \leq \| P \circ|\psi\rangle \|+\sqrt{\frac{q}{2^{\lambda}}}
\end{aligned}
$$

$\rightarrow$ Norm of $\| P \circ|\psi\rangle \|$ increase by at most $\sqrt{\frac{q}{2^{\lambda}}}$ after each query
$\rightarrow$ Norm on bad states after $q$ queries is at most $q \sqrt{\frac{q}{2^{\lambda}}}=\sqrt{\frac{q^{3}}{2^{\lambda}}}$
$\rightarrow$ measuring final state yields bad database (containing collision) with probability at most $\frac{q^{3}}{2^{\lambda}}$

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Fact 4: $\| \mathrm{P} \circ(\mathrm{CPhsO} \circ Q \circ|\psi\rangle)\left\|\leq \sqrt{\frac{q}{2^{\lambda}}}\right\| Q \circ|\psi\rangle \|$
Recall the projection $Q$

$$
|\psi\rangle=\sum_{x, y, z, H} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle \Rightarrow Q|\psi\rangle:=\sum_{x, y \neq 0, z} \sum_{\substack{H \notin \text { Collision } \\ H(x)=\perp}} \alpha_{x, y, z, H}|x, y, z\rangle \otimes|H\rangle
$$

Thus,

$$
\mathrm{CPhsO} \circ Q \circ|\psi\rangle=\sum_{x, y \neq 0, z} \sum_{\substack{H \notin \text { Collision, } \\ H(x)=\perp}} \alpha_{x, y, z, H}|x, y, z\rangle \otimes \sum_{w} 2^{-\lambda / 2}|H \cup(x, w)\rangle
$$

## Proof of Theorem 2

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$$

Thus,

$$
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$$

Where $\operatorname{Bad}(H)$ is the set of ouputs already recorded in $\mathrm{H} . \rightarrow|\operatorname{Bad}(H)| \leq q$

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Theorem 2. For any adversary making $q$ queries to CStO or CPhsO and an arbitrary number of database read queries, if the database $D$ is measured after the $q$ queries, the resulting database will contain a collision with probability at most $O\left(q^{3} / 2^{n}\right)$

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\mathrm{P} \circ \mathrm{CPhsO} \circ Q \circ|\psi\rangle=\sum_{x, y \neq 0, z} \sum_{\substack{H \notin \text { Collision, } \\ H(x)=\perp}} \alpha_{x, y, z, H}|x, y, z\rangle \otimes \sum_{w \in \operatorname{Bad}(H)} 2^{-\lambda / 2}|H \cup(x, w)\rangle
$$

Where $\operatorname{Bad}(H)$ is the set of outputs already recorded in $\mathrm{H} . \rightarrow|\operatorname{Bad}(H)| \leq|H| \leq q$

$$
\begin{aligned}
& \| \mathrm{P} \circ(\mathrm{CPhsO} \circ Q \circ|\psi\rangle) \|^{2}=\sum_{x, y \neq 0, z} \sum_{\substack{H \notin \text { Collision, } \\
H(x)=\perp}} \sum_{w \in \operatorname{Bad}(H)} 2^{-\lambda} \alpha_{x, y, z, H}^{2} \leq q 2^{-\lambda} \sum_{x, y \neq 0, z} \sum_{\substack{H \notin \text { Collision, } \\
H(x)=\perp}} \alpha_{x, y, z, H}^{2} \\
& =q 2^{-\lambda} \| Q \circ|\psi\rangle \|^{2}
\end{aligned}
$$

## Course Evaluation

- I would appreciate if you take the time to fill out your course evaluation.
- What did you like about the course? What could be improved? Let me know!
- Your feedback is anonymous and will not impact grades.


## Final Project Presentation (Next Week)

| Tuesday, April 25 | Thursday, April $27^{\text {th }}$ |
| :--- | :--- |
| Albert Yu (3:00-3:18 PM) | Blake and Xiuyu (3:00-3:28 PM) |
| Nicolas Harrell (3:18-3:36 PM) | Adithya and Jacob (3:28-3:56 PM) |
| Hongoa Wang (3:36-3:54 PM) | Jimmy Hwang (3:56-4:14 PM) |
| Zhongtang Luo (3:54-4:12 PM) |  |

Individuals: 14 minute presentation +3 minute $Q \& A+1$ minute transition
Groups: 24 minute presentation +3 minute $Q \& A+1$ minute transition

- It is expected that both team members will give part of the presentation
- You may choose how to divide the presentation

E-mail slides to Hassan at least one hour before class on the day of your presentation (CC me)

## Final Exam and Project Report

- Final Exam (Take Home):
- Released Thursday, April $27^{\text {th }}$ at 5PM
- Due: Friday, April $28^{\text {th }}$ at 5PM on Gradescope
- Should take $\approx 2$ hours
- Project Report
- 8-12 pages
- Introduce/Motivate the Problem, Define the Problem Clearly, Summarize Related Work et...
- Results
- Note: This can include failed approaches if you clearly describe what you tried and explain why this approach did not work.
- Future Work: If you were to continue working on this problem what would you do?
- Official Due Date: Friday, April $28^{\text {th }} @ 11: 59 P M$
- E-mail me a PDF and CC Hassan
- I won't penalize late solutions submitted before Thursday, May $4^{\text {th }}$ at 11:59PM $\odot$


## Side Channel: Memory Access Pattern

```
Program 1: secret value \(x\)
if \((x<5)\)
    \(z=A[0]\)
    \(\mathrm{A}[1]=\mathrm{z}^{*} \mathrm{z}\)
else
    \(\mathrm{z}=\mathrm{A}[100]\)
    \(\mathrm{A}[101]=\mathrm{z}^{*} \mathrm{z}\)
```

Suppose Attacker Learns Memory Access pattern was (100, 101).

What can attacker conclude?

## Side Channel: Memory Access Pattern

## When could attacker learn memory access pattern?

Scenario 1: User is storing (encrypted) array on an untrusted server

Scenario 2: Program 1 (trusted) is running on the same machine as program 2 (untrusted)

- Operating System ensures that program 2 cannot access program 1's memory. What is the problem?


## Side Channel: Memory Access Pattern

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RAM

|  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $A[0]$ | $A[1]$ | $\ldots$ | $A[100]$ | $A[101]$ | $Y$ | $\ldots$ |
| $B[0]$ | $B[1]$ | $\ldots$ | $B[100]$ | $B[101]$ | $Z$ |  |
|  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |

Cache


Executing P1

## Side Channel: Memory Access Pattern

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RAM

|  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $A[0]$ | $A[1]$ | $\ldots$. | $A[100]$ | $A[101]$ | $Y$ | $\ldots$ |
| $B[0]$ | $B[1]$ | $\ldots$. | $B[100]$ | $B[101]$ | $Z$ |  |
|  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |

Cache

|  |  | Load A[101] |
| :--- | :--- | :--- |
| $A[100]$ |  |  |
| $B[100]$ |  |  |
|  |  |  |
|  |  |  |
|  |  |  |

Executing P1

## Side Channel: Memory Access Pattern

## When could attacker learn memory access pattern?

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RAM

|  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $A[0]$ | $A[1]$ | $\ldots$. | $A[100]$ | $A[101]$ | $Y$ | $\ldots$ |
| $B[0]$ | $B[1]$ | $\ldots$ | $B[100]$ | $B[101]$ | $Z$ |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ |  |  |  |
|  |  |  |  |  |  |  |

Cache


Executing P1

CPU

## Side Channel: Memory Access Pattern

## When could attacker learn memory access pattern?

Scenario 2: Program 1 (trusted) is running on the same machine as program 2 (untrusted)

- Operating System ensures that program 2 cannot access program 1's memory. What is the problem?

RAM

|  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $A[0]$ | $A[1]$ | $\ldots$. | $A[100]$ | $A[101]$ | $Y$ | $\ldots$ |
| $B[0]$ | $B[1]$ | $\ldots$. | $B[100]$ | $B[101]$ | $Z$ |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ |  |  |  |

Cache

| $A[101]$ |  | Load $B[100]$ |  |
| :--- | :--- | :---: | :---: |
| $A[100]$ | $A P U$ |  |  |
| $B[100]$ | $B[101]$ | $\xrightarrow{\text { In-Cache Already } \rightarrow}$Quick response |  |
| $\ldots$ | $\ldots$ |  |  |

## Side Channel: Memory Access Pattern

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|  |  |  |  |  |  |  |
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| $A[0]$ | $A[1]$ | $\ldots$. | $A[100]$ | $A[101]$ | $Y$ | $\ldots$ |
| $B[0]$ | $B[1]$ | $\ldots$. | $B[100]$ | $B[101]$ | $Z$ |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ | $\ldots$ |  |  |
|  |  |  | $\ldots$ |  |  |  |

Cache


## ORAM

Slides: Adapted from TCC 2022 Test of Time Talk
"A Walk in the ORAM Forest: About oblivious RAMs and something about trees" (Jesper Buus Nielsen)

## Definition: ORAM = ObliviousArray

- An ORAM implements an array of N words of $\log (N)$ bits
- Operations on array: operations
- Operations on server memory: probes
- $\mathbf{O}=01,0_{2}, \ldots, O_{n}$
- oi $\in\{$ (read, p), (write, p, v) \}
- $A\left(\mathrm{o}_{\mathrm{i}}\right)=$ list of server positions probed
- $A(0)=\left(A\left(O_{1}\right), A\left(O_{2}\right), \ldots, A\left(O_{n}\right)\right)$
- Oblivious: $\left|\mathbf{0}^{1}\right|=\left|\mathbf{0}^{2}\right| \Rightarrow \Delta\left(\mathrm{A}\left(\mathbf{0}^{1}\right), \mathrm{A}\left(\mathbf{o}^{2}\right)\right) \leq \varepsilon$
- Perfect: $\varepsilon=0$
- Statistical: $\varepsilon=$ negl.
- Computational: | $\Delta \mid=$ poly
- Online: Simulate one operation at a time
- Offline: Gets o up front and can plan

- Amortised overhead: $\lim _{n}|A(0)| /|0|$
- Worst-case overhead: $\max _{\mathrm{i}}\left|\mathrm{A}\left(\mathrm{o}_{\mathrm{i}}\right)\right|$


## Genesis



- Oded Goldreich: Towards a Theory of Software Protection and Simulation by Oblivious RAMs. STOC 1987
- "We show how to implement $n$ fetch instructions to a [RAM] memory of size $m$ by making less than $n \cdot m^{\varepsilon}$ actual accesses, for every fixed $\varepsilon>0$."
- Rafail Ostrovsky: An Efficient Software Protection Scheme. CRYPTO 1989
- Poly-logarithmic overhead
- Oded Goldreich, Rafail Ostrovsky: Software Protection and Simulation on Oblivious

RAMs. J. ACM 43(3). 1996

- Logarithmic lower bound in balls-in-bins model


## Trivial ORAM

- Probe entire memory on each operation
- Perfect
- Overhead = N



## Trivial ORAM

- Alice sends $c_{1}=\operatorname{Enc}_{\mathrm{K}}\left(x_{1}\right), \ldots ., c_{n}=\operatorname{Enc}_{\mathrm{K}}\left(x_{n}\right)$ to Bob for storage
- When Alice wants to load/write $x_{i}$ she requests for Bob to send all ciphertexts.
- Subtle problem for write operations:
- What if Alice sends back $\left(c_{1}^{\prime}=\operatorname{Enc}_{K}\left(x_{1}{ }^{\prime}\right), c_{2}, \ldots c_{n}\right)$ ?
$\rightarrow$ Bob learns that Alice was writing to location 1.
- Solution: Alice generates fresh ciphertexts $c_{i}^{\prime}=\operatorname{Enc}_{\mathrm{K}}\left(x_{i}{ }^{\prime}\right)$ for every (even if $x_{i}^{\prime}=x_{i}$ is unchanged)


## Trivial ORAM

- Alice sends $c_{1}=\operatorname{Enc}_{\mathrm{K}}\left(x_{1}\right), \ldots ., c_{n}=\operatorname{Enc}_{\mathrm{K}}\left(x_{n}\right)$ to Bob for storage
- In remainder of this lecture we will assume that items are encrypted and that Alice always remembers to re-encrypt files every time it is touched.
- We can now focus only on the memory access pattern (i.e., probes).
- Memory Access Pattern: $\mathbf{0}=01,0_{2}, \ldots, O_{n}$
- oi $\boldsymbol{\epsilon}\{$ (read, p), (write, p, v) \}
- Length of access pattern $|\mathbf{0}|=\mathbf{n}$


## Trivial ORAM

- Memory Access Pattern: $\mathbf{0}=0_{1}, 0_{2}, \ldots, O_{n}$
- oi $\boldsymbol{\epsilon}\{$ (read, p), (write, p, v) \}
- Length of access pattern $|\mathbf{0}|=\mathbf{n}$
- A(0) memory access pattern (probes) induced by ORAM compiler
- Note: A(0) is a random variable
- Trivial ORAM:
- $A\left(O_{i}\right)=(1, \ldots, n)$
- $A(0)=\left(A\left(0_{1}\right), \ldots, A\left(O_{n}\right)\right)=((1, \ldots, n), \ldots,(1, \ldots, n))$


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## ORAM Security/Limitation

- Memory Access Pattern: $\mathbf{0}=0_{1}, 0_{2}, \ldots, O_{n}$
- oi $\mathcal{E}\{$ (read, p ), (write, $\mathrm{p}, \mathrm{v}$ ) $\}$
- Length of access pattern $|\mathbf{0}|=\mathbf{n}$
- Security Guarantee: For any two access patterns $\left|\mathbf{O}^{\mathbf{1}}\right|=\mid \mathbf{O}^{\mathbf{2}}$ of the same length an attacker cannot distinguish between $A\left(\mathbf{0}^{\mathbf{1}}\right)$ and $A\left(\mathbf{O}^{\mathbf{2}}\right)$
- Identical Distributions/Statistically Close/Computationally Indistinguishable
- Limitation: If $\left|\mathbf{0}^{1}\right|>\left|\mathbf{0}^{2}\right|$ there are no guarantees.
- Can append $\left|\mathbf{0}^{2}\right|$ with dummy operations
- Must append to maximum running time
- Efficiency bottleneck.


## Oblivious Shuffling

- Initial Array: $\mathrm{A}[0], \ldots, \mathrm{A}[\mathrm{n}-1]$
- Shuffled Array: $A[\pi(0)], \ldots, A[\pi(n-1)]$
- Security Goals:
- $\pi$ is random permutation
- Attacker who observes memory access pattern during shuffling cannot distinguish between $\pi$ and $\pi^{\prime}$ (random unrelated permutation)
- Can do this using $\tilde{O}(N)$ probes


## Oblivious Shuffling

- Initial Array: A[0],...,A[n-1]
- Shuffled Array: $A[\pi(0)], \ldots, A[\pi(n-1)]$
- Shuffle via Merge Sort?
- Suppose we have obliviously shuffled $\mathrm{L}=A[0], \ldots, A\left[\frac{n}{2}-1\right]$ and $\mathrm{R}=$ $A\left[\frac{n}{2}\right], \ldots, A[n-1]$
- Obtained $\pi_{L}(\mathrm{~L})=A\left[\pi_{L}(0)\right], \ldots, A\left[\pi_{L}(n-1)\right]$ and $\pi_{R}(R)=$ $A\left[\pi_{R}\left(\frac{n}{2}\right)\right], \ldots, A\left[\pi_{R}(n-1)\right]$
- Merge with $\tilde{O}(N)$ probes?
- Trickier than it seems...


## Oblivious Merge (Attempt 1)

$\operatorname{Merge}\left(i, A\left[\pi_{L}(0)\right], \ldots, A\left[\pi_{L}(n-1)\right], j, A\left[\pi_{R}\left(\frac{n}{2}\right)\right], \ldots, A\left[\pi_{R}(n-1)\right]\right)$
Set $b=\left\{\begin{array}{ll}0 & \text { w. } p \frac{|L|-i}{|L|-i+|R|-j} \\ 1 & \text { otherwise }\end{array}\right.$ and $Y=\left\{\begin{array}{c}A\left[\pi_{L}(i)\right] \text { if } b=0 \\ A\left[\pi_{R}\left(\frac{n}{2}+j\right)\right] \text { otherwise }\end{array}\right.$
$\mathrm{Z}:=\operatorname{Merge}\left(i+1-b, A\left[\pi_{L}(0)\right], \ldots, A\left[\pi_{L}(n-1)\right],|R|-b, A\left[\pi_{R}\left(\frac{n}{2}\right)\right], \ldots, A\left[\pi_{R}(n-1)\right]\right)$
Return $\mathrm{Y} \circ \mathrm{Z}$

Initial Run with $\mathbf{i = 0}, \mathbf{j}=\mathbf{0}$. Problem?

## Oblivious Merge (Attempt 1)

$$
\begin{aligned}
& \operatorname{Merge}\left(i, A\left[\pi_{L}(0)\right], \ldots, A\left[\pi_{L}(n-1)\right], j, A\left[\pi_{R}\left(\frac{n}{2}\right)\right], \ldots, A\left[\pi_{R}(n-1)\right]\right) \\
& \text { Set } b= \begin{cases}0 & \text { w.p } \frac{|L|-i}{|L|-i+|R|-j} \text { and } Y=\left\{\begin{array}{c}
A\left[\pi_{L}(i)\right] \text { if } b=0 \\
A\left[\pi_{R}\left(\frac{n}{2}+j\right)\right] \text { otherwise }
\end{array}\right. \\
\mathrm{Z}:=\operatorname{Merge}\left(i+1-b, A\left[\pi_{L}(0)\right], \ldots, A\left[\pi_{L}(n-1)\right],|R|-b, A\left[\pi_{R}\left(\frac{n}{2}\right)\right], \ldots, A\left[\pi_{R}(n-1)\right]\right)\end{cases}
\end{aligned}
$$

Return $\mathrm{Y} \circ \mathrm{Z}$
Initial Run with $\mathbf{i = 0 , j = 0}$. Problem? Red probes leak information about the final permutation!

## Permute and Guard

- Guard of size G
- Store array + G dummy elements uniformly random permuted on server in PM
- Write: Write to Guard
- Read
- First read from Guard
- If not found, read from PM
- If found read dummy from PM
- Write to Guard
- Refresh: After G operations
- Join Guard and PM and create new fresh PM with fresh $K$



## Permute and Guard

- Computational
- Price of $G$ operations:
- Operations: $G^{\star} \mid G u a r d l \approx G^{2}$
- Permute $\approx N$
- Balances at $G=\sqrt{N}$
- Amortised overhead: $A O H \approx \sqrt{N}$



## Recursive ${ }^{\text {eg }}$

- Use a larger Guard implemented as a Map data structure on a smaller ORAM
- We can do ORAM with $\mathrm{AOH}=N^{1 / 2}$
- Set the smaller ORAM to size $\approx N^{2 B}$ and capacity $N^{2 ß}$ in the map
- Price of $G$ operations:
- Op.s: $G^{*} O H_{\text {Guard }}=G\left(N^{2 \beta}\right)^{1 / 2}=G N^{1 /}$
- Permute $\approx N$
- Balances at $G \approx N / N^{1 / \beta} \approx N^{2 / B}$
- Amortised overhead: $\mathrm{AOH} \approx N^{2} N^{2 / 3}=N^{1 / 3}$



## Recursive

- Use a larger Guard implemented as a Map data structure on a smaller ORAM
- We can do ORAM with $\mathrm{AOH}=N^{1 /(k-1)}$
- Set the smaller ORAM to size $\approx N^{k-1) k}$ and (write, $p, v$ ) capacity $N^{(k-1) / k}$
- Price of $G$ operations:
- Op.s: $G^{\star} O H_{\text {Guard }}=G N^{1 k}$
- Permute $\approx N$
- Balances at $G \approx N / N^{1 k} \approx M^{k-1)} k$
- Amortised overhead: $\mathrm{AOH} \approx N^{\left(N^{k-1}\right) k}=N^{1 / k}$



## Hierarchical

- After each $2^{\prime}$ operations shuffle the elements in levels $1, \ldots$, $i$ and store them in level $i$ using fresh $\mathrm{PRF}_{K}$
- Lemma: All guards get emptied before they are full
- Some nitty-gritty stuff with dummies and enough room for the map data structure and dummies



## AOH of Hierarchical

- Guard $i$ has size about $2^{i} i$
- Guard $i$ is sorted every $2^{-i}$ operations at a price of $2^{i}{ }^{2}$
- Guard $i$ costs amortised $i^{2}$ per operation
- $1+2^{2} \ldots+\log (N)^{2} \approx \log (N)^{3}$
- $\mathrm{AOH}: \log (N)^{3}$



## Ball in Bins

- Lower bound [GO'96]: To implement $N$ operations you must make $N \log (N)$ probes (if you use a balls-in-bins ORAM)
- Balls-in-Bins:
- The ORAM construction does not look at the data being stored, it is treated atomically
- Can only confuse the adversary by swapping balls around



## Now What？

－By 1996 some very basic question are open：

Is computational security inherent？


銯Is amortisation inherent or can we get poly－log worst－case overhead？
素Can we go below $\mathrm{OH} \log (N)$ using a non－balls－in－bins ORAM？
教Is the $\mathrm{OH} \log (N)$ or $\log (N)^{3}$ or somewhere in between？


## March 2010: Information Theoretic Solutions

- Ivan Damgård, Sigurd Meldgaard, Jesper Buus Nielsen: Perfectly Secure Oblivious RAM without Random Oracles. TCC 2011
- Posted on IACR ePrint on 2 March 2010
- AOH: $\log \left(N^{3}\right.$
- Perfect security
- Also:
- Method for ORAM in MPC
- $N \log (N)$ lower bound on amount of randomness ORAM must store
- Milkós Ajtai. Oblivious RAMs without cryptographic assumptions. STOC 2010
- Posted on ECCC on 6 March 2010
- STOC had deadline November 52009
- Poly-logarithmic overhead, constant in exponent not explicitly given
- Statistical error for $t$ operations: $t^{\text {tog(t) }}$


## 2011: Worst-Case Poly-Log OH

- Elaine Shi, T.-H. Hubert Chan, Emil Stefanov, Mingfei Li: Oblivious RAM with O((log N $)^{3}$ ) Worst-Case Cost. ASIACRYPT 2011
- Worst case $\log (N)^{3}$
- Amortised overhead $\log (N)^{2}$ when using DMN'11 as a tool: ORAM on small internal buckets
- Statistical
- Michael T. Goodrich, Michael Mitzenmacher, Olga Ohrimenko, Roberto Tamassia: Oblivious RAM simulation with efficient worstcase access overhead. CCSW 2011
- Worst case $N^{1 / 2 l o g(N)^{2}}$
- Worst case $\log (N)^{2}$ with large client memory.
- Computational
- Eyal Kushilevitz, Steve Lu, Rafail Ostrovsky: On the (in)security of hash-based oblivious RAM and a new balancing scheme. SODA 2012
- IACR ePrint August 2011
- Worst case $\log (N)^{2} / \log (\log (N))$
- Computational
- The last two paper uses "deamortisation":
- Have two hash maps on each level
- Shuffle one while using the other and swap when the shuffling is done


## 2011-2013: Path ORAM

- Elaine Shi, T.-H. Hubert Chan, Emil Stefanov, Mingfei Li: Oblivious RAM with O((log N)³) Worst-Case Cost. ASIACRYPT 2011
- Worst case overhead $\log (N)^{3}$
- Emil Stefanov, Marten van Dijk, Elaine Shi, Christopher W. Fletcher, Ling Ren, Xiangyao Yu, Srinivas Devadas: Path ORAM: an extremely simple oblivious RAM protocol. CCS 2013
- Worst case overhead $\log (N)^{2}$
- PathORAM Idea:
- Fix the tree (do not shuffle the levels as Ajtai and DMN'11)
- Store $V_{p}$ on path to uniformly random leaf $L_{p}$
- Use smaller ORAM $L$ with size $N^{\prime}=N / 2$ and word size $w=2 \log (N)$
- For all $p^{\prime}=p$ div 2 store $\left(L_{2 p+0,} L_{2 p+1}\right)$ in $L[p]$


## Path ORAM

- Divide layer $i$ into $2^{i}$ buckets of size $O(1)$
- Impose a BST on the buckets
- Do not shuffle the levels!
- Assign position $p$ of array to a uniformly random leaf $L_{p}$ of the tree - Invariant: $\left(L_{p}, V_{p}\right)$ is always to be found in bucket on path to $L_{p}$
- Inject fresh ( $L_{p}, V_{p}$ ) at root after access
- Read/write by r/w entire path to $L_{p}$
- Push ( $L, V$ ) pairs as low as possible before writing the path back
- Use a stash to store overflow from buckets
- Worst-case OH: $\log (N)$



## Its ORAMs All the Way Down!

- Recursively store position maps for levels of size $N_{i}$ in ORAMs of size $N_{i-1}=N_{i} / 2$
- Statistical security
- Worst-case OH: $\log (N)^{2}$
- $\sum_{i=1, \ldots, \log (N)} \log \left(2^{i}\right) \approx \log (N)^{2}$
- If the word size is $w=\log (N)^{2}$ then the OH becomes $\log (N)$
- Not trivial
- Not unreasonable in practice



## 2016-2018: What About that Lower Bound?!

- Elette Boyle, Moni Naor: Is There an Oblivious RAM Lower Bound? ITCS 2016
- Points out the following about the Goldreich-Ostrovsky lower bound:
- It only applies to "balls-in-bins" algorithms, i.e., algorithms where the ORAM may only shuffle stored values around and not apply any sophisticated encoding of the data
- It only applies to computationally unbounded adversaries
- But it applies even to off-line algorithms and improving it will involve switching to considering on-line or proving unconditional lower bounds of circuits for sorting
- Kasper Green Larsen, Jesper Buus Nielsen: Yes, There is an Oblivious RAM Lower Bound! CRYPTO 2018
- Applies to all types of on-line algorithms
- Applies also to computationally bounded adversaries


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- Applies to all types of on-line algorithms
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- Mihai Patrascu, Erik D. Demaine: Logarithmic Lower Bounds in the Cell-Probe Model. SIAM
J. Comput. 35(4) 2006
- Introduced the "Information transfer" technique
- On-line algorithms turns time into location by putting events on a time-line
- Reasoning about how information moves around in space is much, much easier than reasoning about computational complexity
- Put a binary tree on top to reason about how information is moved
- LN'18: The "Information transfer" technique normally does not apply to array maintenance but when combined with obliviousness suddenly it does


## PanORAMa

After the Cuckoo:
Lookup is log( $N$ )
Amortised OH is $\log (N)^{2}$ because of having to shuffle the levels


Sarvar Patel, Giuseppe Persiano, Mariana Raykova, Kevin Yeo: PanORAMa: Oblivious RAM with Logarithmic Overhead. FOCS 2018

Amortised Overhead: $\log (N) \log (\log (N))$
No need to shuffle merged levels:
The remaining, untouched elements are already randomly permuted!
Extract the untouched elements
One can do this in $\mathrm{O}(N \log (\log (N)))$
Sorting small buckets of size $\mathrm{O}(\log (\mathrm{N})$
Randomly merge the permuted untouched elements
Only has to add $O(M$ randomness but suffers $\log (\log (N)$ to do it obliviously

## OptORAMa

Gilad Asharov, Ilan Komargodski, Wei-Kai Lin, Kartik Nayak, Enoch Peserico, Elaine Shi: OptORAMa: Optimal Oblivious RAM. EUROCRYPT 2020

Amortised Overhead: O(log N)
Has O(M) oblivious, deterministic tight compaction aloorithm!
Tight compactions: Sort elements marked 0 or 1 such that all marked 0 appear first
Circumvents 0-1 lower bound by doing non-comparison operations
Extract the unused elements using tight compaction
Merge-shuffle: Just a "reverse tight compaction" which is O(M)
Paper is 73 pages so I must have simplified somewhere:-)

## All at Once!?!?!

- Perfect, worst-case, $\mathrm{OH}=\log (N)$ ?
- Michael A. Raskin, Mark Simkin: Perfectly Secure Oblivious RAM with Sublinear Bandwidth Overhead. ASIACRYPT 2019
- Worst case $\mathrm{OH}=\sqrt{ } \mathrm{N}$
- Computational, worst-case, $\mathrm{OH}=\log (\mathrm{N})$ ?
- Gilad Asharov, Ilan Komargodski, Wei-Kai Lin, Elaine Shi: Oblivious RAM with Worst-Case Logarithmic Overhead. CRYPTO 2021
- New deamortisation technique compatible with merge-shuffle and compaction
- Perfect, amortised, $\mathrm{OH}=\log (\mathrm{N})$ ?
- T.-H. Hubert Chan, Elaine Shi, Wei-Kai Lin, Kartik Nayak: Perfectly Oblivious (Parallel) RAM Revisited, and Improved Constructions. ITC 2021
- $\mathrm{AOH}=\mathrm{O}\left(\log (N)^{3} / \log (\log (N))\right)$
- Perfect, worst-case, $\mathrm{OH}=\log (N)$ ?
- Statistical, worst-case, $\mathrm{OH}=\log (N)$
- Path ORAM for word-size $\log (N)^{2}$


## ORAMs with Special Properties

- ORAMs good for MPC
- Marcel Keller, Peter Scholl: Efficient, Oblivious Data Structures for MPC. ASIACRYPT 2014
- Xiao Wang, T.-H. Hubert Chan, Elaine Shi: Circuit ORAM: On Tightness of the Goldreich-Ostrovsky Lower Bound. CCS 2015
- Parallel ORAM
- Elette Boyle, Kai-Min Chung, Rafael Pass: Oblivious Parallel RAM and Applications. TCC 2016
- Round Complexity
- David Cash, Andrew Drucker, Alexander Hoover: A Lower Bound for One-Round Oblivious RAM. TCC 2020
- $\sqrt{ } \mathrm{N}$ Overhead
- Oh a $\sqrt{ }$ again!?


## - Random-index ORAM

- Shai Halevi, Eyal Kushilevitz: Random-Index Oblivious RAM. TCC Yesterday.
- This one is One-Round...


## Other Oblivious Data Structures

- Xiao Shaun Wang, Kartik Nayak, Chang Liu, T.-H. Hubert Chan, Elaine Shi, Emil Stefanov, Yan Huang: Oblivious Data Structures. CCS 2014
- Riko Jacob, Kasper Green Larsen, Jesper Buus Nielsen: Lower Bounds for Oblivious Data Structures. SODA 2019
- $\Omega(\log N)$ lower bounds for oblivious stacks, queues, deques, priority queues and search trees
- Giuseppe Persiano, Kevin Yeo: Lower Bounds for Differentially Private RAMs. EUROCRYPT 2019
- Constant DP security of a single operation implies $\Omega(\log N) \mathrm{OH}$
- Information transfer does not work here, introduces chronogram technique to ORAM
- Fredman, M., Saks, M.: The cell probe complexity of dynamic data structures. STOC 1989
- Kasper Green Larsen, Mark Simkin, Kevin Yeo: Lower Bounds for Multi-server Oblivious RAMs. TCC 2020
- K servers of which the adversary can see the access pattern to only one
- If better than approx $1 / K$ security then $\mathrm{OH} \Omega(\log N)$
- Zahra Jafargholi, Kasper Green Larsen, Mark Simkin: Optimal Oblivious Priority Queues. SODA 2021
- $\mathrm{OH}=10 \log (N)$
- Ilan Komargodski, Elaine Shi: Differentially Oblivious Turing Machines. ITCS 2021 - OH O( $\log \log N)$
- Differentially private stack can be done with $\mathrm{OH} \mathrm{O}(\log \log N)$.

Thanks for Listening


## How to Record Quantum Queries

and Applications to Quantum Indifferentiability

Mark Zhandry

Princeton University \& NTT Research



## The (Classical) Random Oracle Model (ROM)



## The (Classical) Random Oracle Model (ROM)

[Bellare-Rogaway'93]


## Typical ROM Proof: On-the-fly Simulation



## Typical ROM Proof: On-the-fly Simulation

Allows us to:

- Know the inputs adversary cares about
- Know the corresponding outputs
- (Adaptively) program the outputs
- Easy analysis of bad events (e.g. collisions)


## The Quantum Random Oracle Model (QROM)



## Problem with Classical Proofs in QROM



# Problem with Classical Proofs in QROM 

## Observer Effect: <br> Learning anything about quantum system disturbs it

$\frac{\text { H. }}{\text { (r) }}$ answers obliviously, so no disturbance

Reduction must answer obliviously, too?

## Typical QROM Proof



H fixed once and for all at beginning

## Limitations

Allows us to:

- Know the inputs adversary cares about?
- Know the corresponding outputs?
- (Adaptively) program the outputs?
- Easy analysis of bad events (e.g. collisions)?


## Limitations

Allows us to:

- INnow the inputs adversary cares about? $X$
- 皆now the corresponding outputs?
-(Adaptively) program the outputs?
/ $/ X$
- Easy analysis of bad ovents (e.g. collicions)? X


## Limitations

Good News: Numerous positive results (30+ papers)

Bad News: Still some major holdouts
Indifferentiable domain extension

Fiat-Shamir

Luby-Rackoff

ROM è ICM

## Example: Domain Extension for Random Oracles

Q: Does Merkle-Damgård preserve random oracle-ness?


## Example: Domain Extension for Random Oracles

A: Yes(ish) [Coron-Dodis-Malinaud-Puniya’05]
How? Indifferentiability [Maurer-Renner-Holenstein’04]


## Quantum Indifferentiability?

Concurrently considered by [Carstens-Ebrahimi-Tabia-Unruh'18]


## Quantum Indifferentiability?



This Work:

# On-the-fly simulation of quantum random oracles 

(aka Compressed Oracles)

## Step 1: Quantum-ify (aka Purify)

- Quantum-ifying (aka purifying) random oracle:



## Step 1: Superposition of Oracles

Initial oracle state:


Oracle's state


Adversary's query

## Step 2: Look at Fourier Domain



## Step 2: Look at Fourier Domain



## Step 3: Compress

## Observation:

After $\mathbf{q}$ queries, $\hat{\mathbf{H}}$ is non-zero on at most $\mathbf{q}$ points


## Step 3: Compress

Initial oracle state: $\}$
Query (x, y, $\left.D^{\wedge}\right)$ :
(1) If $\nexists\left(x, y^{\prime}\right) \in{ }^{\wedge} D: \mathcal{D}=D^{\wedge}+(x, 0)$
(2) Replace $\left(x, y^{\prime}\right) \in \wedge^{\wedge} \mathbf{D}$ with ( $\mathbf{x}, \mathbf{y}^{\prime} \oplus \mathbf{y}$ )
(3) If $(\mathbf{x}, \mathbf{0}) \in^{\wedge} \mathbf{D}$ : remove it


## Step 4: Revert back to Primal Domain



## Step 4: Revert back to Primal Domain



## Compressed Oracles

Allows us to:

- Know the inputs adversary cares about?
- Know the corresponding outputs?
- (Adaptively) program the outputs? $x$
Fixed by [Don-Fehr-Majenz-Schaffner'19,Liu-Z'19], later this session!
- Easy analysis of bad events (e.g. collisions)? $\downarrow$


## So, what happened?

## Recall...

## Observer Effect:

Learning anything about quantum system disturbs it


Compressed oracles decode such disturbance

## Caveats

Outputs in database $\neq \mathbf{0}$ in Fourier domain
$\Rightarrow y$ values aren't exactly query outputs

## Examining $\mathbf{x , y}$ values perturbs state

$\Rightarrow$ Still must be careful about how we use them

But, still good enough for many applications...

## Applications In This Work

## Quantum Indiff. of <br> Merkle-Damgård

## Easily re-prove quantum lower bounds: $\Omega\left(\mathbf{N}^{1 / 2}\right)$ queries needed for Grover search $\Omega\left(\mathbf{N}^{1 / 3}\right)$ queries needed for collision finding $\Omega\left(\mathbf{N}^{1 /(k+1)}\right)$ queries needed for $\mathbf{k}$-SUM

CCA-security of plain
Fujisaki-Okamoto

## Further Applications

[Alagic-Majenz-Russell-Song'18]:
Quantum-secure signature separation
[Liu-Z'19a]: Tight bounds
for multi-collision problem
[Liu-Z'19b]: Fiat-Shamir
([Don-Fehr-Majenz-Schaffner'19]: direct proof)
[Czajkowski-Majenz-Schaffner-Zur'19]:
Indifferentiability of Sponge
[Chiesa-Manohar-Spooner'19]:
[Hosoyamada-Iwata'19]:
zk-SNARKs
[Bindel-Hamburg-Hülsing-Persichetti' 19]:
Tighter CCA security proofs

## Lessons Learned



Always purify your oracles!

