
On Perfect and Adaptive Security in Exposure-Resilient Cryptography

Yevgeniy Dodis, New York University

Amit Sahai, Princeton

Adam Smith, MIT

Problem: Partial Key Exposure

- Alice needs to store a cryptographic **key**
- She wants to store her key on a hard drive
- Eve may break in and access some **limited number of bits**
- How can Alice store her key so it remains secure?

Problem: Partial Key Exposure

- Standard cryptography:
 - No guarantees even if only **tiny** fraction of key is leaked
- Paradigm: “Exposure-Resilient Cryptography”
 - Build primitives that remain secure even when **most** of the key is leaked.

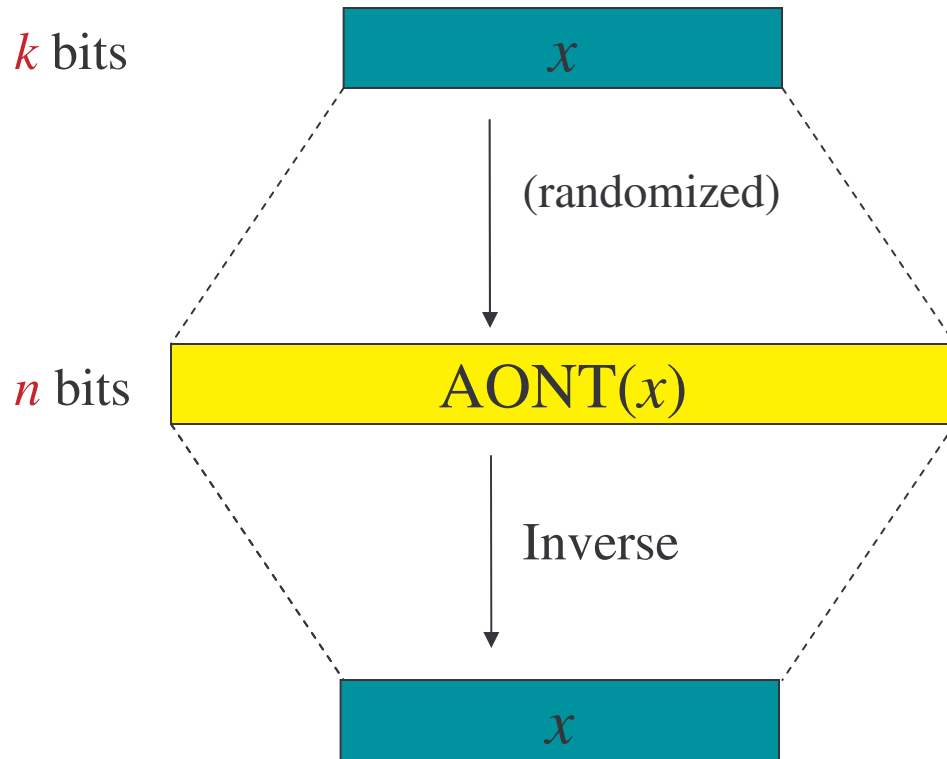
“Exposure-Resilient Cryptography”

- Two main primitives:
 - All-or-Nothing Transforms (AONT)
 - Exposure-Resilient Functions (ERF)
- This talk focuses on AONT.
- ERF also discussed in the paper.

All-or-Nothing Transform (AONT)

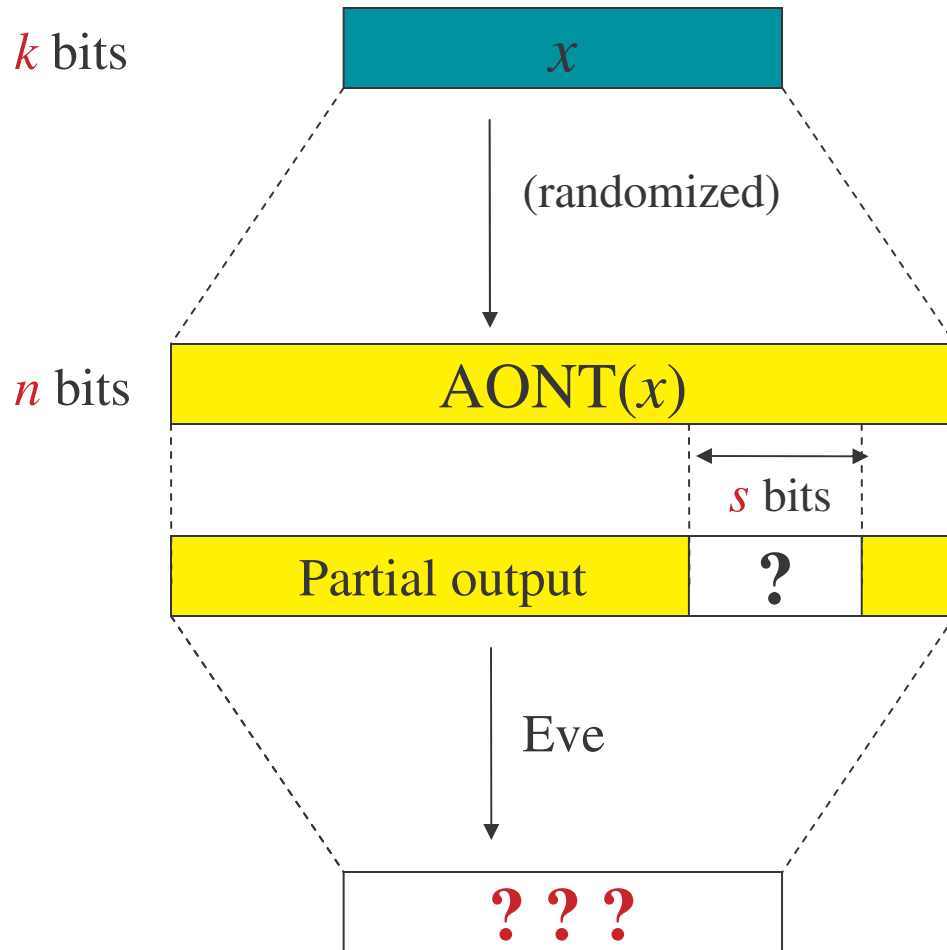
- Randomized encoding
- No **key**
- Increases length
- If you know:
 - **All** of the output..... **Recover** the **whole** input
 - **Part** of the output..... **No information** about input

All-or-Nothing Transform (AONT)



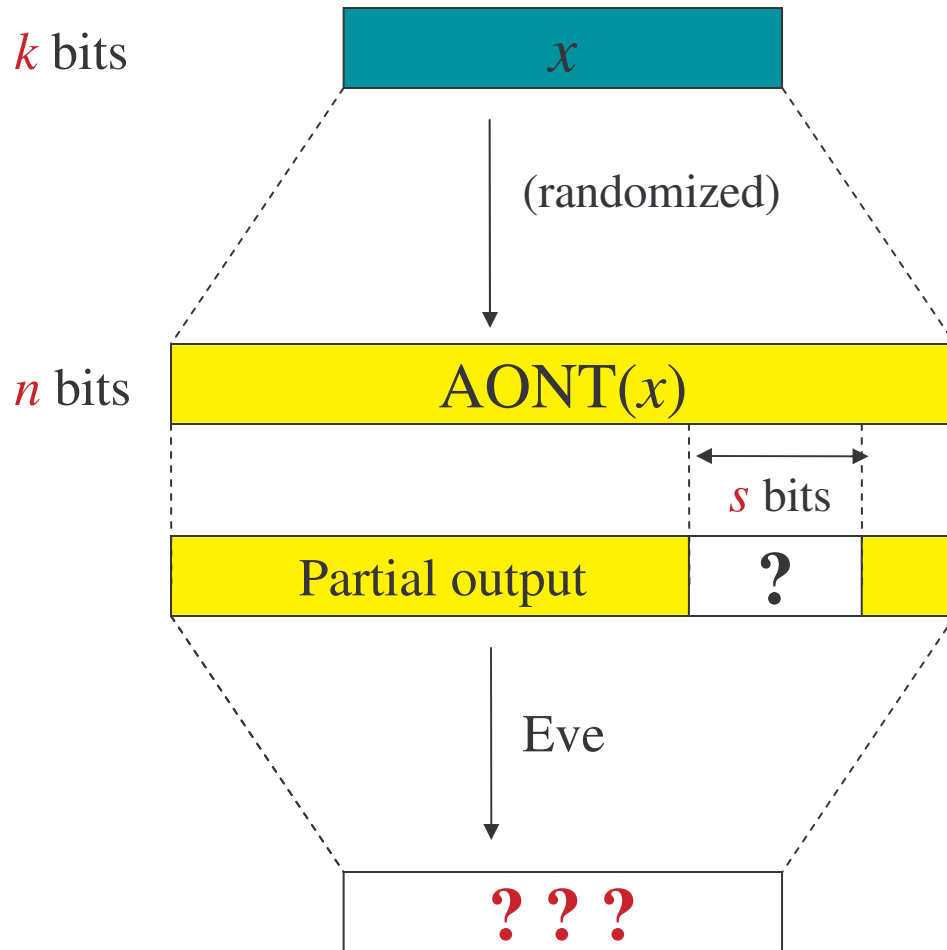
- Randomized encoding
- With all of the output
can recover input

All-or-Nothing Transform (AONT)



- Randomized encoding
- With all of the output can recover input
- If missing s bits of output, **No Information** about input

All-or-Nothing Transform (AONT)

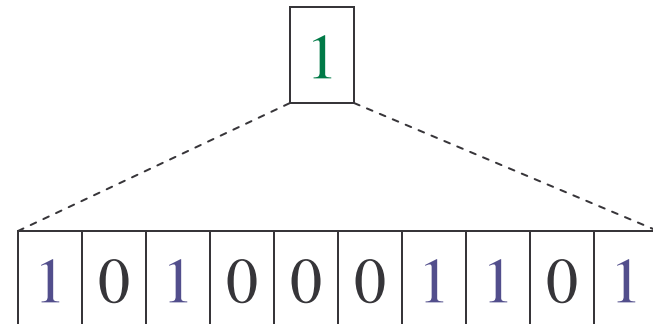


Special case of
“Ramp Secret Sharing”

- Shares are 1 bit each
- All n shares
⇒ find x
- $\leq n-s$ shares
⇒ no information

Example of AONT

- Simplest example: parity
- Encode a single bit



- $\text{AONT}(b) = \text{random string } y \text{ such that } \text{PARITY}(y)=b$
 - Given **all** of y , computing b is easy
 - If Eve misses **any bit** of y , then no information on b
 - $k = 1$
 - $s = 1$

Problem: Partial Key Exposure

- Alice needs to store a cryptographic **key** (say for signing or for encryption)
- She wants to store her key on a hard drive
- Eve may break in and access some **limited number of bits** on the hard drive.
- How can Alice store her key so it remains secure?

Solution: All-or-Nothing Transform

Alice stores **encoding** $AONT(x)$
instead of her **key** x

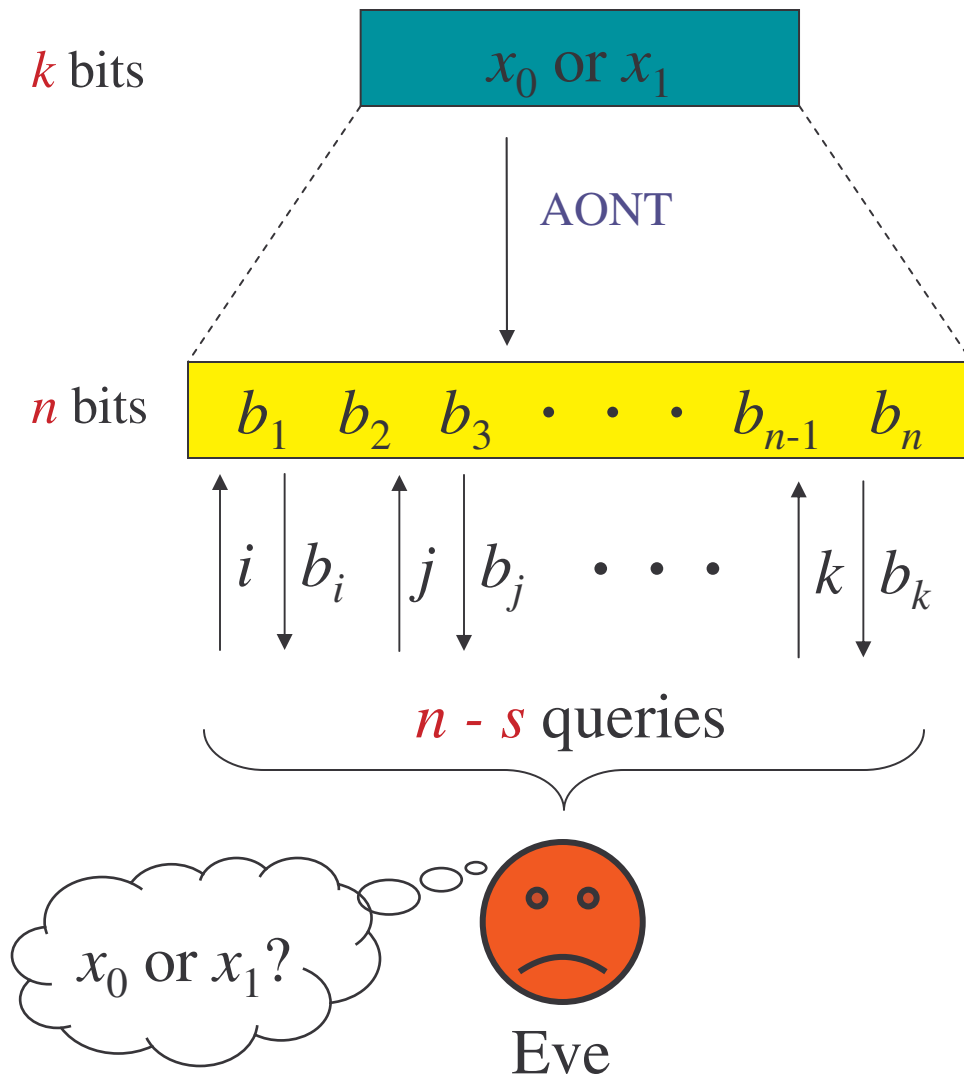
- Other applications:
 - Introduced by Rivest for strengthening block ciphers
 - Many recent works with constructions and applications:

Rivest '97, Boyko '99, Desai '00, Matyas et al. '96, Jakobson et al. '99,
Blaze '96, Bellare-Boldyreva '00, Shin-Rhee '99, Canetti et al. '00, ...

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - **How well** can we do against adaptive adversaries?
(i.e. what parameters can we achieve?)

How to define adaptive security?



- Similar to semantic security for cryptosystems
- Secret is one of two possibilities
 x_0 or x_1
- Eve queries output **one bit at a time**
- Eve tries to guess whether she saw $\text{AONT}(x_0)$ or $\text{AONT}(x_1)$
- Success probability should be $\approx 1/2$

Strengths of secrecy

- **Perfect:**

- secret **independent** of adversary's view
- i.e. distribution on views **the same** for any two secrets

Success prob. = $1/2$

- **Statistical:**

- distributions on views are **statistically close**
for any two secrets

Success prob. = $1/2 \pm \epsilon$

- **Computational:**

- distributions on views are **computationally indistinguishable**
for any two secrets

Success prob. = $1/2 \pm \epsilon$
when Eve is poly-time

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - **How well** can we do against adaptive adversaries?
(i.e. what parameters can we achieve?)

Are Previous Constructions Secure ?

- Previous constructions..... under Adaptive adversaries
 - Perfect secrecy: Secure, but bad parameters
 - Statistical secrecy: Insecure
 - Standard computational secrecy: Insecure
 - Random oracle/cipher model: Secure, but must assume random oracle
- [CDHKS '00]
- [Boyko '99, Rivest '97, Desai '00,...]

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - How well can we do against adaptive adversaries?
(i.e. what parameters can we achieve?)

Main results

- **Perfect** secrecy:
 - Static security = adaptive security
 - New **lower bound** for AONT
 - Can't reveal more than half of output! (when secret $> \log n$ bits)
- **Statistical** secrecy:
 - Previous constructions **insecure** against adaptive adversary
 - Simple **near-optimal, probabilistic** constructions of:
 - Almost-perfect resilient Functions (APRF)
 - All-or-Nothing Transform (AONT)
 - Exposure-resilient Functions (ERF)
- **Computational** secrecy:
 - Combine **statistical secrecy** with pseudo-random generator [CDHKS '00]
 - (Almost) arbitrary parameters ($s \ll k$)


$$s \approx k$$

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - How well can we do against adaptive adversaries?
(i.e. **what parameters** can we achieve?)
 - Lower bound for **perfect** secrecy
 - Near-optimal **statistical** constructions

Lower bound on perfect AONT

Lower bound on perfect AONT

- Recall AONT encodes k bits into n bits
 - Know all n bits of output \Rightarrow Recover whole input
 - Know $n - s$ bits \Rightarrow No information

- We show:

$$s \geq \frac{n}{2} + \left(1 - \frac{n}{2 \cdot (2^k - 1)} \right)$$

- In particular:

When $k > \log n$ we must hide at least $1/2$ of the the output!

Lower bound on perfect AONT

- Main ideas:
 - View $\{0,1\}^n$ as a graph (a.k.a. “the hypercube”)
 - Perfect AONT are “balanced, weighted colorings” of hypercube
 - Use Fourier analysis over graph $\{0,1\}^n$ (*i.e.* over group Z_2^n)
 - Details in the paper.
- Previously proven only for a related primitive:
Resilient Functions [Friedman ‘92, Bierbrauer et al ‘96]
- Generalizes technique of Friedman ‘92
 - Previous bound is a special case

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - How well can we do against adaptive adversaries?
(i.e. what parameters can we achieve?)
 - Lower bound for **perfect** secrecy
 - Near-optimal **statistical** constructions

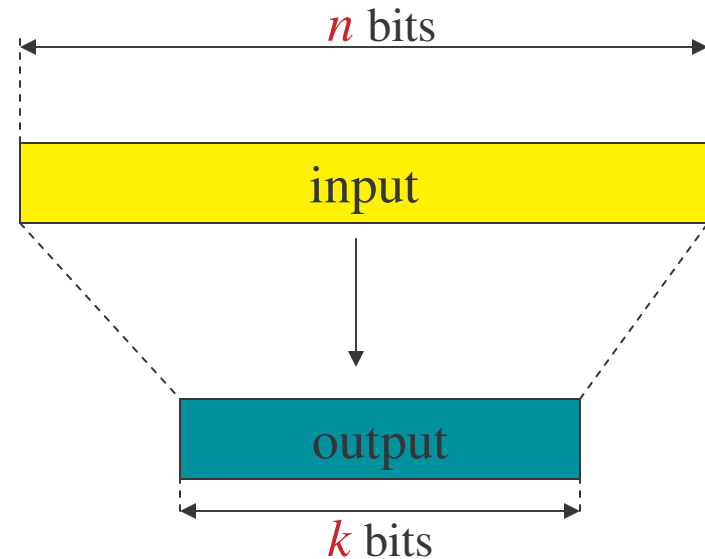
Statistical Security against adaptive adversaries

How to achieve adaptive statistical security

- Reduce adaptive security of AONT to a **static property**
- **Main tool:**
 - “Almost-perfect resilient functions” (APRF) [Kurosawa-Johansson-Stinson ‘97]
- Simple construction of APRF: Pick **random hash function**
 - Will be an APRF with high probability
 - Near-optimal parameters
- Good APRF’s imply good AONT

Main tool: Resilient Functions

- Functions from n bits to k bits
- Intuition:
 - Choosing all but s bits of input gives **no information** on output.
- Various formalizations of this in statistical setting (see paper).
- Strongest notion is APRF (“almost-perfect resilient function”)
 - Eve can **fix** $n-s$ bits of input
 - Remaining s bits chosen **at random**
 - Distribution on output still close to uniform **on all points**



Main tool: Resilient Functions

- Functions from n bits to k bits

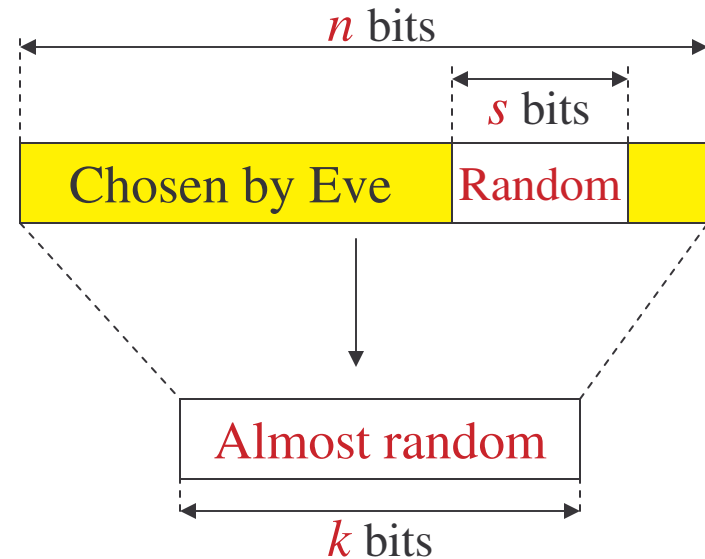
- Intuition:

- Choosing all but s bits of input gives **no information** on output.

- Various formalizations of this in statistical setting (see paper).

- Strongest notion is APRF (“almost-perfect resilient function”)

- Eve can **fix** $n-s$ bits of input
- Remaining s bits chosen **at random**
- Distribution on output still close to uniform **on all points**



How to construct APRF?

- Kurosawa et al. '97:
 - Coding-theoretic construction
 - Achieves $s \geq \frac{n+k}{2}$
 - Very far from trivial bound of $s \geq k$
- Intuition: A random function is an APRF with high probability (proof by Chernoff bounds)
- Construction: Pick a random $(n/\log n)$ -wise independent hash function
 - $s = k + 2\log(1/\epsilon) + O(\log n)$ with high probability
 - $= k + o(k)$ when $k \gg \log n$
 - Proof relies on tail bounds on n -wise independent variables
 - Also used in Trevisan-Vadhan'00 for constructing deterministic extractors

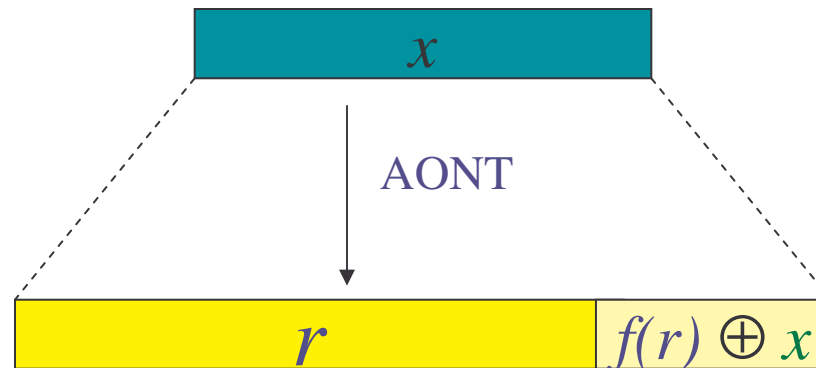
APRF \Rightarrow AONT: One-time pad

- Say f is an APRF with output length n
- $\text{AONT}(x) = (r, f(r) \oplus x)$ for random string r

$$- n' = n + k$$

$$- s' = s$$

$$- k' = k$$



- Still get $s' \approx k'$
- Slight expansion of output
- Note: If f is efficiently invertible, no pad is necessary

This work: Adaptive Security

- In many situations, Eve might:
 - Learn bits **one at a time**
 - **Choose** which bit to learn next based on
 - what she has seen so far
 - any **partial information** she has about the input
- **Questions:**
 - How to **define** adaptive security?
 - Are **previous constructions** secure ?
 - How well can we do against adaptive adversaries?
(i.e. what parameters can we achieve?)
 - Lower bound for **perfect** secrecy
 - Near-optimal **statistical** constructions

Conclusions

- Considered **adaptive security** for exposure-resilient primitives
- **Perfect** setting
 - New lower bound for AONT, matches previous bounds for Exposure-Resilient Functions
- **Statistical** setting
 - Reduce to construction of Almost-Perfect Resilient Functions
 - Near-optimal constructions of APRF, ERF, AONT