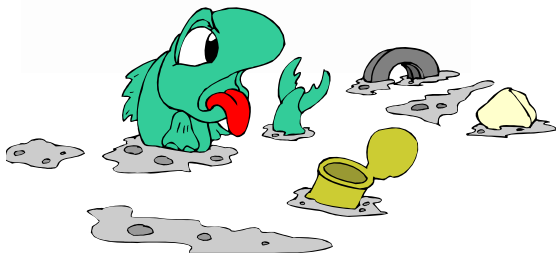


KEY DERIVATION

WITHOUT

ENTROPY WASTE

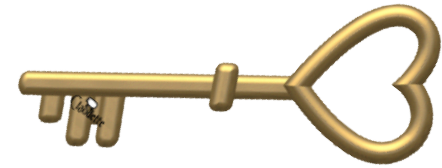


Yevgeniy Dodis

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Joint work with Krzysztof Pietrzak and Daniel Wichs

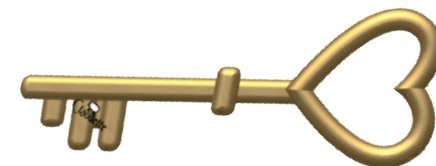
Key Derivation



2

- Setting: application P needs m -bit secret key R
- Theory: pick *uniformly random* $R \leftarrow \{0,1\}^m$
- Practice: have "imperfect randomness" $X \in \{0,1\}^n$
 - physical sources, biometric data, partial key leakage, extracting from group elements (DH key exchange), ...
- Need a "bridge": *key derivation function (KDF)*
 $h: \{0,1\}^n \rightarrow \{0,1\}^m$ s.t. $R = h(X)$ is "good" for P
 - ... only assuming X has "minimal entropy" k

Formalizing the Problem

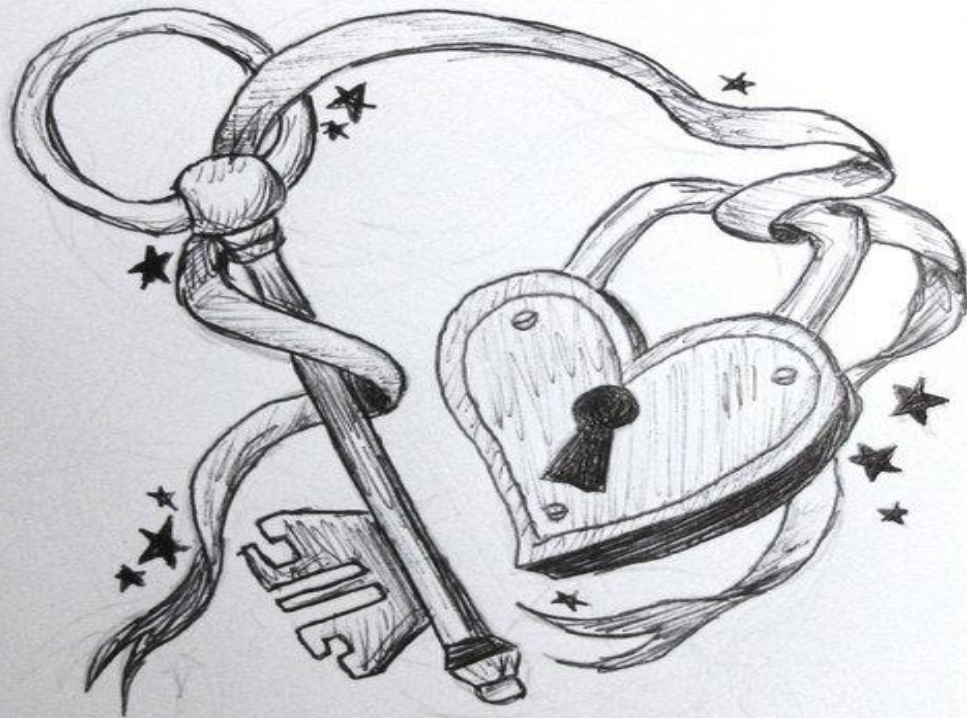


3

- Ideal Model: pick uniform $R \leftarrow U_m$ as the key
 - Assume P is ε -secure (against certain class of attackers A)
- Real Model: use $R = h(X)$ as the key, where

Real Security ε' \approx Ideal Security ε

- Goal: minimize k s.t. P is 2ε -secure using $R = h(X)$
 - Equivalently, minimize entropy loss $L = k - m$
 - (If possible, get information-theoretic security)
 - Note: we design h but must work for any (n, k) -source X



□ Note: we design h but must work for any (n, k) -source X

Old Approach: Extractors

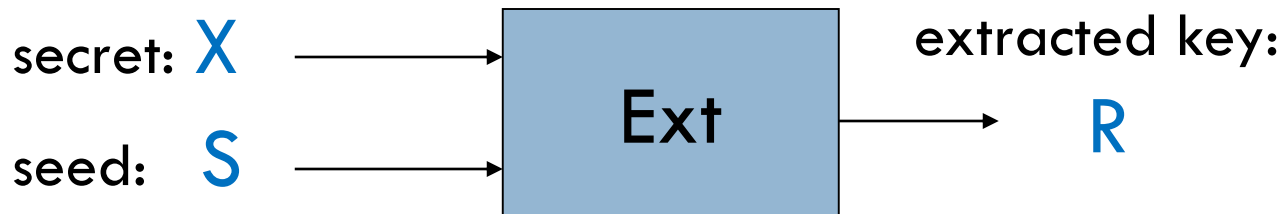


5

- Tool: Randomness Extractor [NZ96].
 - Input: a weak secret X and a uniformly random seed S .
 - Output: extracted key $R = \text{Ext}(X; S)$.
 - R is uniformly random, even conditioned on the seed S .

$$(\text{Ext}(X; S), S) \approx (\text{Uniform}, S)$$

- **Many uses in complexity theory and cryptography.**
 - Well beyond key derivation (de-randomization, etc.)



Old Approach: Extractors



6

- Tool: Randomness Extractor [NZ96].
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 - R is uniformly random, even conditioned on the seed S .

$$(\text{Ext}(X; S), S) \approx (\text{Uniform}, S)$$

- (k, ε) -extractor: given any secret (n, k) -source X , outputs m secret bits “ ε -fooling” any distinguisher D :

statistical distance

$$| \Pr[D(\text{Ext}(X; S), S) = 1] - \Pr[D(U_m, S) = 1] | \leq \varepsilon$$

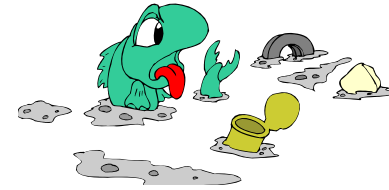
Extractors as KDFs



7

- Lemma: for **any** ε -secure P needing an m -bit key, (k, ε) -extractor is a KDF yielding security $\varepsilon' \leq 2\varepsilon$
- LHL [HILL]: universal hash functions are (k, ε) -extractors where $k = m + 2\log(1/\varepsilon)$
- Corollary: For **any** P , can get entropy loss $2\log(1/\varepsilon)$

How Bad is $2\log(1/\varepsilon)$ Entropy Loss?



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- Many sources do not have “extra” $2\log(1/\varepsilon)$ bits
 - ▣ Biometrics, physical sources, DH keys on elliptic curves
 - DH: low $k \Rightarrow$ smaller group size \Rightarrow higher efficiency
 - ▣ AES-based P: $\varepsilon = 2^{-64}$, $m = 128 \Rightarrow k = 256 = 2m$ ☹
- Heuristic extractors have “no entropy loss”: $k = m$
- End Result: practitioners prefer heuristic key derivation to provable key derivation [DGH⁺,Kra]
- Can we **provably** match it, despite RT-bound?

Extractors as KDFs



9

- Lemma: for **any** ε -secure P needing an m -bit key, (k, ε) -extractor is a KDF yielding security $\varepsilon' \leq 2\varepsilon$
- LHL [HILL]: universal hash functions are (k, ε) -extractors where $k = m + 2\log(1/\varepsilon)$
- Corollary: For **any** P , can get entropy loss $2\log(1/\varepsilon)$
- RT-bound [RT]: for any extractor, $k \geq m + 2\log(1/\varepsilon)$
 - entropy loss $2\log(1/\varepsilon)$ seems necessary ☹️

□ ... or is it?



Side-Stepping RT



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- Do we need to derive **statistically random R** ?
 - **Yes** for one-time pad ☹️...
 - **No** for many (most?) other applications **P** 😊 !

this work

Series of works “beating” RT [BDK⁺11, DRV12, DY13, DPW14]

For the first time match heuristic extractors!

Punch line: Difference between
Extraction and **Key Derivation** !

New Approach/Plan of Attack



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- **Step1.** Identify **general class** of applications **P** which work “well” with any **high-entropy key** R
 - Interesting in its own right !

- **Step2.** Build good **condenser**: relaxation of extractor producing **high-entropy** (but *non-uniform!*) derived key $R = h(X)$

Plan of Attack



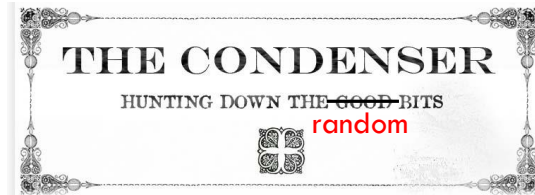
14

✓ **Step 1.** Argue *any* unpredictability applic. \mathcal{P} works well with (only) a **high-entropy key** R

▣ $H_\infty(R) \geq m - d \Rightarrow \epsilon' \leq 2^d \cdot \epsilon$

▣ **Step 2.** Build good **condenser**: relaxation of extractor producing **high-entropy** (but *non-uniform!*) derived key $R = h(X)$

Randomness Condensers



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- (k, d, ε) -**condenser**: given (n, k) -source X , outputs m bits R “ ε -close” to some $(m, m-d)$ -source Y :

$$(\mathbf{Cond}(X; S), S) \approx_{\varepsilon} (Y, S) \quad \text{and} \quad H_{\infty}(Y | S) \geq m - d$$

- **Cond** + Step1 $\Rightarrow \varepsilon' \leq (1 + 2^d) \cdot \varepsilon$

- Extractors: $d = 0$ but only for $k \geq m + 2\log(1/\varepsilon)$ ☹️

- **Our Main Result**: $d = 1$ with $k = m + \log\log(1/\varepsilon) + 4$

- KDF: $\log(1/\varepsilon)$ -independent hash function works!

Balls and Bins



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- Reduces to simple balls-and-bins game:
 - Throw 2^k balls into 2^m bins
 - Pick a random ball x
 - Lose if $|Bin(x)| > 2^d$.
- Goal: given $d, m, \varepsilon \Rightarrow \min_{S \subseteq [k]}$. $\Pr[Lose] \leq \varepsilon$
- Easy calculation \Rightarrow parameters of theorem if throw balls totally independently
- Observation: $\log(1/\varepsilon)$ -independence suffices!

improve $|S|$ to $O(n \log k)$ using “gradual increase of independence” [CRSW11]

Unpredictability Extractors



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□ **Corollary:** provably secure KDF with entropy loss $\log\log(1/\varepsilon) + 4$ for all unpredictability applications

□ Implicitly built $(k, \varepsilon, \varepsilon')$ -unpredictability extractors:

$$\Pr[\mathbf{D}(U_m, S) = 1] \leq \varepsilon \Rightarrow \Pr[\mathbf{D}(\mathbf{UExt}(X; S), S) = 1] \leq \varepsilon'$$

▪ got $\varepsilon' = 3\varepsilon$ and $k = m + \log\log(1/\varepsilon) + 4$

Example: CBC-MAC, $\varepsilon = 2^{-64}$, $m = 128$

LHL: $k = 256$; **Now:** $k = 138 \cong 128$ (RO)

Unpredictability Extractors



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□ **Corollary:** provably secure KDF with entropy loss $\log\log(1/\varepsilon) + 4$ for all unpredictability applications

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▪ got $\varepsilon' = 3\varepsilon$ and $k = m + \log\log(1/\varepsilon) + 4$

□ More generally, $\varepsilon' = \varepsilon \cdot (1 + \log(1/\varepsilon)) \cdot 2^{m-k}$

▪ E.g., $\varepsilon' = \varepsilon \cdot (1 + \log(1/\varepsilon))$ when $k = m$

▪ CBC-MAC: $k = m = 128 \Rightarrow \varepsilon = 2^{-57.9}$ (vs. 2^{-63} RO)

Options for Avoiding RT



20

- ✓ Route 1: *implicitly* restrict D by considering **special classes of applications** P [BDK⁺11, DRV12, DY13, DPW14]
 - ▣ This paper: $L = \log\log(1/\epsilon)$ for all unpredictability P
 - ▣ [BDK⁺11, DY13]: $L = \log(1/\epsilon)$ for all “square-friendly” P
(includes unpred. P , but also CPA enc, weak PRF, ...)
- ▣ Route 2: efficiently samplable sources X [DGKM12]

Efficient Samplability?



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- **Theorem** [DPW14]: efficient samplability of X does not help to improve entropy loss below
 - $2\log(1/\varepsilon)$ for **all** applications P (RT-bound)
 - Affirmatively resolves “SRT-conjecture” from [DGKM12]
 - $\log(1/\varepsilon)$ for all **square-friendly** applications P
 - $\log\log(1/\varepsilon)$ for all **unpredictability** applications P

Options for Avoiding RT



22

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 - ▣ This paper: $L = \log\log(1/\epsilon)$ for all unpredictability P
 - ▣ [BDK⁺11, DY13]: $L = \log(1/\epsilon)$ for all “square-friendly” P
(includes unpred. P , but also CPA enc, weak PRF, ...)
- ✓ Route 2: efficiently samplable sources X [DGKM12]
- ▣ Route 3: computat. bounded D (**pseudo-randomness**)

Computational Assumptions?



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- **Theorem** [DGKM12, DPW14]: ~~SRT conjecture~~ \Rightarrow efficient **Ext** beating RT-bound for all computationally bounded $D \Rightarrow$ **OWFs exist**
- How far can we go with OWFs/PRGs/...?
 - One of the main open problems
- **Current Best** [DY13]: “computational” **extractor** with entropy loss $2\log(1/\varepsilon) - \log(1/\varepsilon_{\text{prg}})$
 - “Computational” **condenser**?

Summary



- Difference between **extraction** and **KDF**
 - ▣ $\log\log(1/\epsilon)$ loss for all **unpredictability** apps
 - ▣ $\log(1/\epsilon)$ loss for all **square-friendly** apps
(+ motivation to study “**square security**”)
- Efficient samplability does **not** help ☹️
- Good **computational** extractors require **OWFs** ☹️
- **Main challenge**: better “computational” KDFs

Questions?



