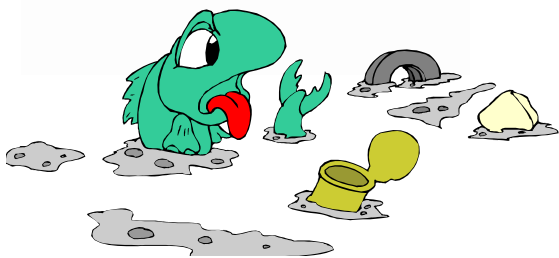


*KEY DERIVATION*

*WITHOUT*

*ENTROPY WASTE*



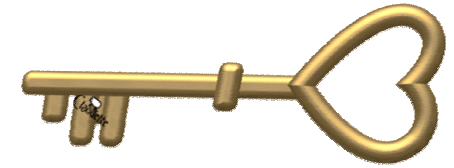
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Yevgeniy Dodis

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Based on joint works with B. Barak, H. Krawczyk, O. Pereira, K. Pietrzak, F-X. Standaert, D. Wichs and Y. Yu

# 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"  $k$

# Dreaming Big

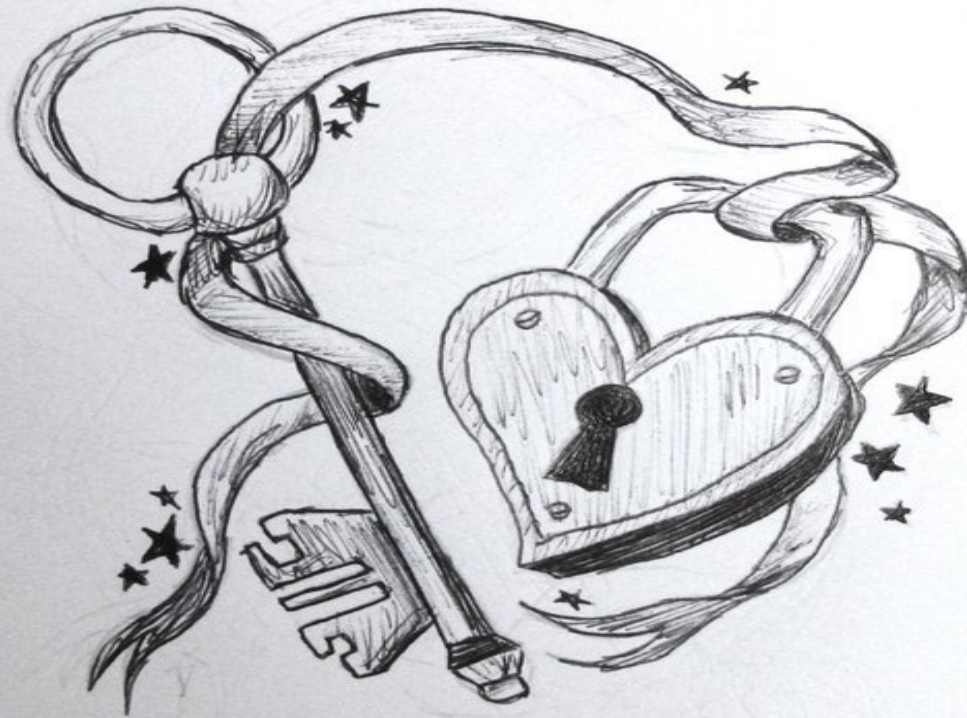


3

- Question 1: minimal entropy  $k$  enough to achieve “real security”  $\approx$  “ideal security” for P?
  - Dream 1: can get  $k \approx m$  (no “entropy loss”) !

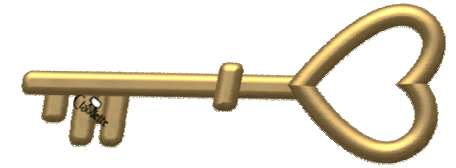
Question/Dream 3: can we ever hope to achieve comparable security without entropy loss ?!

- Question 2: best security degradation when  $k \approx m$  ?
  - Dream 2: (almost) no security degradation !



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

# Formalizing the Problem



5

- Ideal Model: pick *uniform*  $R \leftarrow U_m$  as the key
  - Assume  $P$  is  $\epsilon$ -secure against certain class of attackers  $A$
- Real Model: use  $R = h(X)$  as the key, where
  - $\text{min-entropy}(X) = H_\infty(X) \geq k$  ( $\Pr[X = x] \leq 2^{-k}$ , for all  $x$ )
  - $h: \{0,1\}^n \rightarrow \{0,1\}^m$  is a (carefully designed) KDF
- Goal: prove that  $P$  is  $\epsilon'$ -secure in the **real** model (against same/similar class of attackers  $A$ )
  - Note: **we** design  $h$  but must work for **any**  $(n, k)$ -source  $X$
- What is the smallest  $\epsilon'$ ???

# Dreaming Big, formally



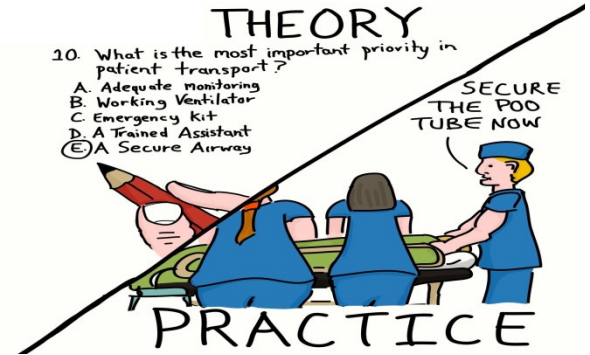
6

- Question 1: minimal  $k$  (call it  $k^*$ ) to get  $\varepsilon' = 2\varepsilon$ ?
  - Dream 1: can get  $k^* \approx m$  (no “entropy loss”) !

Question/Dream 3: can we ever hope to achieve  $\varepsilon' = O(\varepsilon)$  security when  $k \approx m$  (no entropy loss) ?!

- Question 2: smallest  $\varepsilon'$  (call it  $\varepsilon^*$ ) when  $k = m$  ?
  - Dream 2: can get  $\varepsilon^* = O(\varepsilon)$  (no security degradation) !

# Theory vs. Practice



- Practice: heuristic key derivation ( $h = \text{SHA, MD5, ...}$ )
  - common belief among practitioner: **Dream 3 is TRUE!**
- *Amazing* (heuristic) bound in “random oracle” model:

$$\varepsilon' \leq \varepsilon + \varepsilon \cdot 2^{m-k}$$

- “implies”  $\varepsilon^* = 2\varepsilon$  and  $k^* = m$  **at the same time!**
- Despite lack of “practical” attacks, lots of (valid) criticism [DHK<sup>+</sup>04, Kra10, BDK<sup>+</sup>11]
- **How close can we come in theory (and practice 😊)?**

# Extractors

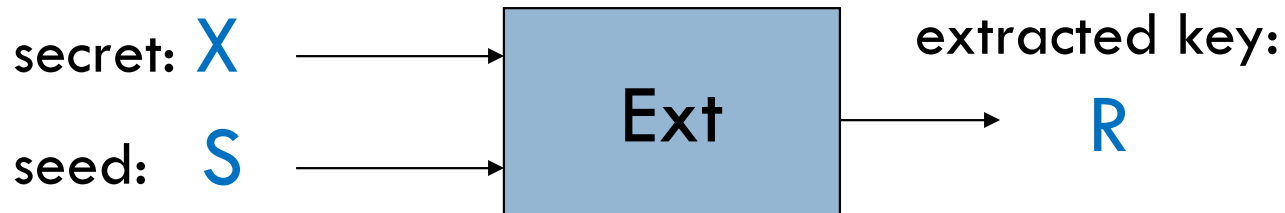


8

- 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.)







# (Seeded) Extractors

9

## □ 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  $S$ .

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

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

statistical distance

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

# Extractors as KDFs

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- Lemma: for **any**  $\epsilon$ -secure  $P$  needing an  $m$ -bit key,  $(k, \delta)$ -extractor is a KDF yielding security  $\epsilon' \leq \epsilon + \delta$
- Note: use potentially **restricted** distinguishers  $D$ 
  - $D$  = combination of attacker  $A$  and challenger  $C$
  - $D$  outputs  $1 \iff A$  “won” (e.g., forged signature) against  $C$
- Best tradeoff between  $m$ ,  $k$  &  $\delta$  in a  $(k, \delta)$ -extractor?

# Leftover Hash Lemma



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- LHL [HILL]: universal hash functions are  $(k, \delta)$ -extractors where  $\delta = \sqrt{2^{m-k}}$
- Corollary: For any  $P$ ,  $\epsilon' \leq \epsilon + \sqrt{2^{m-k}}$ . In particular,
  - $k^* = m + 2\log(1/\epsilon)$  ( entropy loss  $2\log(1/\epsilon)$  enough )
  - $\epsilon^* = 1$  ( no meaningful security when  $k = m$  ☹ )
- RT-bound [RT]: Any  $(k, \delta)$ -extractor  $\Rightarrow \delta \geq \sqrt{2^{m-k}}$ 
  - Above bounds are optimal (in this level of generality) ☹

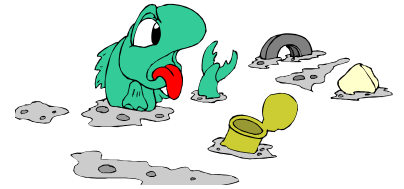
# Theory vs. Practice:

12



Application <b>P</b>	KDF $h$	Sec. Loss $\epsilon' - \epsilon$	$\epsilon^*$ <small>(<math>k=m</math>)</small>	Entr. Loss $k^* - m$	Provable?
Computat. Secure	SHA/RO	$\epsilon \cdot 2^{m-k}$	$2\epsilon$	0	no
<b>ANY</b>	universal hash	$\sqrt{2^{m-k}}$	1	$2\log(1/\epsilon)$	yes

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



13

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

# Options for Avoiding RT



14

- Route 1: restrict the power of distinguisher  $D$  or the class of  $(n, k)$ -sources  $X$ 
  - Ex. 1: efficiently samplable sources  $X$  [DGKM12]
  - Ex. 2: computationally bounded  $D$  (pseudo-randomness)
  - Ex. 3: *implicitly* restrict  $D$  by considering special classes of applications  $P$  [BDK<sup>+</sup>11, DRV12, DY13, DPW13]
- Route 2: do we need to derive statist. random  $R$ ?
  - Yes for OTP; No for many (most?) other applications  $P$ !

# Options for Avoiding RT



15

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

□ Ex. 3: *implicitly* restrict **D** by considering **special classes of applications P** [BDK<sup>+</sup>11, DRV12, DY13, DPW13]

□ Route 2: do we need to derive **statist. random R**?

□ **Yes** for OTP; **No** for many (most?) other applications **P**!

# Unpredictability Applications

16



- $\text{Adv}(\mathbf{A}) = \Pr[\mathbf{A} \text{ wins}] = \Pr[\mathbf{D} \text{ out. } 1] \in [0,1]$ 
  - ▣ signatures, MACs, one-way functions, ... (*not* encryption!)
- Case Study: key derivation for signature/MAC
  - ▣ Assume:  $\Pr[\mathbf{A} \text{ forges sig with uniform key}] \leq \epsilon$  (= negl)
  - ▣ Hope:  $\Pr[\mathbf{A} \text{ forges sig with extracted key}] \leq \epsilon'$  ( $\approx \epsilon$ )
- Key Insight: only care about distinguishers  $\mathbf{D}$  which **almost never** succeed on uniform keys ( $\Pr[.] \leq \epsilon$ ) !
  - ▣ E.g., small **multiplicative** security loss is OK now



# Unpredictability Extractors

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□ **UExt** is  $(k, \varepsilon, \varepsilon')$ -unpredictability extractor if

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

□ **Theorem** [DPW13]: efficient  $(k, \varepsilon, \varepsilon')$ -**UExt** with

□ **Option 1:**  $\varepsilon' = 3\varepsilon$  and  $k = m + \log\log(1/\varepsilon) + 4$

□ **Option 2:**  $\varepsilon' = \varepsilon \cdot (1 + \log(1/\varepsilon))$  and  $k = m$

# Plan of Attack

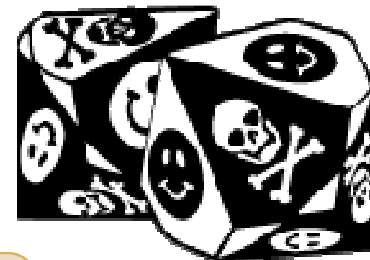


18

- **Step 1.** Argue *any* unpredictability applic.  $P$  works well with (*only*) a *high-entropy key*  $R$ 
  - Of independent interest !

E.g., random  $R$  except first bit  $0 \Rightarrow \epsilon' \leq 2\epsilon$

# 1: Security with Weak Keys



19

Entropy  
deficiency

- Fix  $P$  and any “legal”  $A$
- Let  $f(r) = [\text{Advantage of } A \text{ on key } r] \in [0,1]$
- Ideal Adv.  $\varepsilon = \mathbb{E}[f(U_m)] = \sum_r \frac{1}{2^m} \cdot f(r)$
- Real Adv.  $\varepsilon' = \mathbb{E}[f(R)] = \sum_r p(r) \cdot f(r)$
- **Lemma:** If  $f(r) \geq 0$  and  $H_\infty(R) \geq m - d$ ,  
$$\mathbb{E}[f(R)] \leq 2^d \cdot \mathbb{E}[f(U_m)]$$
- Proof:  $\sum p(r) \cdot f(r) \leq 2^m \cdot \max_r(p(r)) \cdot (\sum \frac{1}{2^m} \cdot f(r))$  ■
- **Corollary:**  $H_\infty(R) \geq m - d \Rightarrow \varepsilon' \leq 2^d \cdot \varepsilon$

# Plan of Attack



20

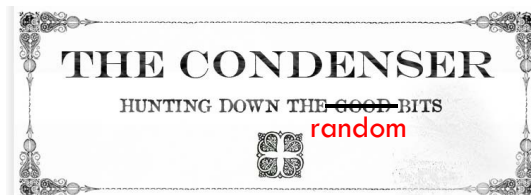
Achieve extremely low  $2^d$  to compose with **Step 1!**

**Option 1:**  $2^d = 2$  and  $k = m + \log\log(1/\epsilon) + 4$

**Option 2:**  $2^d = \log(1/\epsilon)$  and  $k = m$

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

## 2: Randomness Condensers



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

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

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

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



- **Theorem** [DPW13]: efficient  $(k, d, \epsilon)$ -condenser with

- **Option 1:**  $d = 1$  and  $k = m + \log\log(1/\epsilon) + 4$

- **Option 2:**  $d = \log\log(1/\epsilon)$  and  $k = m$

# Balls and Bins



22

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

# Balls and Bins



23

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  - Throw  $2^k$  balls into  $2^m$  bins
  - Pick a random ball  $x$
  - Lose if  $|Bin(x)| > 2^d \cdot 2^{k-r}$
- **Goal:** given  $d, m, \epsilon \Rightarrow \min k$  s.t.  $\Pr[\text{Lose}] \leq \epsilon$
- Easy calculation  $\Rightarrow$  parameters of theorem if throw balls totally independently
- **Observation:**  $\log(1/\epsilon)$ -independence suffices!

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

# Theory vs. Practice:



24

Application <b>P</b>	KDF <i>h</i>	Sec. Loss $\epsilon' - \epsilon$	$\epsilon^*$ <small>(<math>k=m</math>)</small>	Entr. Loss $k^* - m$	Provable?
Computat. Secure	SHA/RO	$\epsilon \cdot 2^{m-k}$	$2\epsilon$	0	no
Unpredict.	$\log(1/\epsilon)$ - wise hash	$\epsilon \cdot \log(1/\epsilon) \cdot 2^{m-k}$	$\epsilon \cdot \log(1/\epsilon)$	$\log \log(1/\epsilon)$	yes
ANY	universal hash	$\sqrt{2^{m-k}}$	1	$2 \log(1/\epsilon)$	yes



# Theory vs. Practice:



25

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

**LHL:**  $\epsilon^* = 1$  and  $k^* = 256$

**Now:**  $\epsilon^* = 2^{-57.9}$  and  $k^* = 138$

**Heuristic:**  $\epsilon^* = 2^{-63}$  and  $k^* = 128$

Sometimes Dreams Come True!



# Indistinguishability Apps?

26

- ❑ Impossible for one-time pad ☹️
- ❑ Still, similar plan of attack:



- ❑ **Step 1.** Identify *sub-class* of indist. applications  $P$  which work well with (*only*) a **high-entropy key  $R$** 
  - Will use **Renyi** entropy instead of min-entropy
  - Weaker inequality, but **still beat LHL**
- ❑ **Step 2.** Build good *condensers* for Renyi entropy

# Simple Inequality



27

- $\mathbf{Col}(R) = \Pr[R_1=R_2] = \sum p(r)^2$ 
  - ▣ Renyi:  $H_2(R) = -\log \mathbf{Col}(R) \geq H_\infty(R)$

- **Lemma:** For **all**  $f$  and  $H_2(R) \geq m-d$ ,

$$|\mathbb{E}[f(R)] - \mathbb{E}[f(U_m)]| \leq \sqrt{2^d - 1} \cdot \sqrt{\mathbb{E}[f(U_m)^2]}$$

- Proof: LHS =  $\left| \sum_r (p(r) - \frac{1}{2^m}) \cdot f(r) \right|$

- CS:  $\leq \sqrt{2^m \sum (p(r) - \frac{1}{2^m})^2} \cdot \sqrt{\frac{1}{2^m} \sum f(r)^2} \dots$  ■

# Why is it Nice?



28

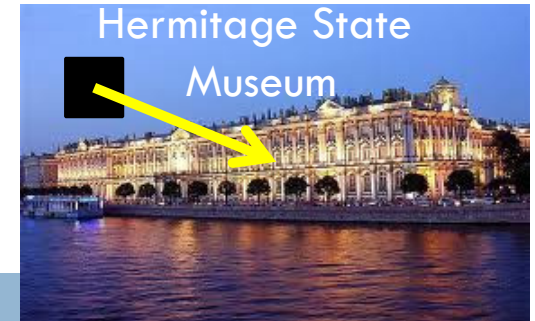
- **Lemma:** For **all**  $f$  and  $H_2(R) \geq m - d$ ,

$$|\mathbb{E}[f(R)] - \mathbb{E}[f(U_m)]| \leq \sqrt{2^d - 1} \cdot \sqrt{\mathbb{E}[f(U_m)^2]}$$

- Works even if  $f(r)$  can be **negative** (indist. OK)
- First term does **not** depend on  $f$  (i.e., appl.  $P$ )
- Second term is for **uniform** distribution
- Nicer entropy for condenser:  $H_2(R) \geq H_\infty(R)$
- **Question:**  $|\mathbb{E}[f(U_m)]| = \varepsilon$ , what is  $\mathbb{E}[f(U_m)^2]$ ?



# Square Security

- Def:  $P$  is  $\sigma$ -square secure (against a class of attackers  $A$ ), if for any  $A \Rightarrow \mathbb{E}[f_A(U_m)^2] \leq \sigma$
- Lemma: If  $P$  is  $\varepsilon$ -secure and  $\sigma$ -square secure, then  $P$  is  $\varepsilon'$ -secure in “ $(m-d)$ -real model”, where  $\varepsilon' \leq \varepsilon + \sqrt{\sigma \cdot (2^d - 1)}$
- Motivates studying square security!
- Question: how does square security  $\sigma$  relate to regular security  $\varepsilon$ ?



# Square-Friendly Applications

30

- $P$  is **square-friendly\*** (SQF) if  $\sigma \leq \varepsilon$
- **Example:** **all unpredictability** applications  $P$ 
  - $f \in [0,1] \Rightarrow \sigma = \mathbb{E}[f^2] \leq \mathbb{E}[f] = \varepsilon$  
- Non-SQF applications: OTP, PRF, PRP, PRG ☹️
- [BDK<sup>+</sup>11, DY13]: **many natural indistinguishability applications are square-friendly !**
-  CPA/CCA-encryption, weak PRFs,  $q$ -wise independent hash functions, ...

\* Allow for small (say, factor of 2) degradation in the efficiency of the attacker  $A$

# Indistinguishability Apps?

31

- Impossible for one-time pad ☹️
- Still, similar plan of attack:



- **Step 1.** Identify *sub-class* of indist. applications  $P$  which work well with (*only*) a **high-entropy key  $R$** 
  - Will use Renyi entropy instead of min-entropy
  - Weaker inequality, but **still beat LHL**
- **Step 2.** Build good *condensers* for Renyi entropy

# Universal Hash Functions



32

- **Universal Hash Family**  $\mathcal{H} = \{ h: \{0,1\}^n \rightarrow \{0,1\}^m \}$ :

$$\forall x \neq x', \Pr_h[ h(x) = h(x') ] = \frac{1}{2^m}$$

- **LHL'**. **Universal** family  $\mathcal{H}$  defines  $(k, d, 0)$ -**condenser**<sub>2</sub> with  $m$ -bit output, where  $2^d - 1 = 2^{m-k}$

- $\Pr[h(X) = h(X')] \leq \Pr[X = X'] + \Pr[h(X) = h(X') \ \& \ X \neq X']$

- $= 2^{d-m} \leq 2^{-k} + 2^{-m}$

■

- **Corollary:** If  $P$  is  $\epsilon$ -secure and square-friendly, then universal hashing yields KDF with  $\epsilon' \leq \epsilon + \sqrt{\epsilon \cdot 2^{m-k}}$



# Theory vs. Practice:



33

Application <b>P</b>	KDF <i>h</i>	Sec. Loss $\epsilon' - \epsilon$	$\epsilon^*$ ( $k=m$ )	Entr. Loss $k^* - m$	Provable?
Computat. Secure	SHA/RO	$\epsilon \cdot 2^{m-k}$	$2\epsilon$	0	no
Unpredict.	$\log(1/\epsilon)$ - wise hash	$\epsilon \cdot \log(1/\epsilon) \cdot 2^{m-k}$	$\epsilon \cdot \log(1/\epsilon)$	$\log \log(1/\epsilon)$	yes
Square- Friendly	universal hash	$\sqrt{\epsilon} \cdot 2^{m-k}$	$\sqrt{\epsilon}$	$\log(1/\epsilon)$	yes
ANY	universal hash	$\sqrt{2^{m-k}}$	1	$2 \log(1/\epsilon)$	yes

# Theory vs. Practice:



34

□ Example: CBC Encryption,  $\epsilon = 2^{-64}$ ,  $m = 128$

**LHL:**  $\epsilon^* = 1$  and  $k^* = 256$

**LHL':**  $\epsilon^* = 2^{-32}$  and  $k^* = 192$

**Heuristic:**  $\epsilon^* = 2^{-63}$  and  $k^* = 128$

# Options for Avoiding RT



35

- Route 1: restrict the power of distinguisher  $D$  or the class of  $(n, k)$ -sources  $X$

□ Ex. 1: efficiently samplable sources  $X$  [DGKM12]

□ Ex. 2: computationally bounded  $D$  (pseudo-randomness)

✓ □ Ex. 3: *implicitly* restrict  $D$  by considering special classes of applications  $P$  [BDK<sup>+</sup>11, DRV12, DY13, DPW13]

- Route 2: do we need to derive **statist. random**  $R$ ?

✓ □ **Yes** for OTP; **No** for many (most?) other applications  $P$ !

# Efficient Samplability

36



- **Theorem** [DPW13]: efficient samplability of  $X$  does not help to improve entropy loss below
  - $2\log(1/\epsilon)$  for **all** applications  $P$  (RT-bound)
    - Affirmatively resolves “SRT-conjecture” from [DGKM12]
  - $\log(1/\epsilon)$  for all **square-friendly** applications  $P$
  - $\log\log(1/\epsilon)$  for all **unpredictability** applications  $P$
- **Idea**: bounded independent  $(n, k)$ -source  $X$  is enough to fool any extractor/condenser/...

# Options for Avoiding RT



37

- Route 1: restrict the power of distinguisher  $D$  or the class of  $(n, k)$ -sources  $X$ 
  - ✓ □ Ex. 1: efficiently samplable sources  $X$  [DGKM12]
  - Ex. 2: computationally bounded  $D$  (pseudo-randomness)
  - Ex. 3: *implicitly* restrict  $D$  by considering special classes of applications  $P$  [BDK<sup>+</sup>11, DRV12, DY13, DPW13]
- Route 2: do we need to derive **statist. random**  $R$ ?
- ✓ □ **Yes** for OTP; **No** for many (most?) other applications  $P$ !



# Minimal Assumptions


38

- **Theorem** [DGKM12, DPW13]: ~~SRT conjecture~~  $\Rightarrow$  efficient **Ext** beating RT-bound for all computationally bounded **D**  $\Rightarrow$  **OWFs exist**
- How far can we go with OWFs/PRGs?
- **Extract-then-Expand** [Kra10]: Beats RT-bound, but only for **medium-to-high** values of  $k$  😞
- **Expand-then-Extract** (aka “dense-model thm”): horrible run-time degradation in reduction 😞

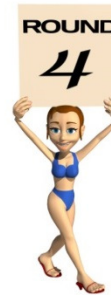
# Computational Extractor



39

- **Idea:** Design **square-friendly** key derivation
  - Good KDF for any **computationally secure P**
- **Solution:** Use **weak** PRF  $f$ : set  $R = f_X(S)$ 
  - wPRF: secure for random (but public) inputs
- **Note:**  $f$  only needs security against **2 queries!**
  - [DY13]: Can easily build using one PRG call:  
“expand-then-extract **w/o time degradation**”!
  - New alternative to “**dense model**” theorem 

# Theory vs. Practice:



40

Application P	KDF h	Sec. Loss $\epsilon' - \epsilon$	$\epsilon^*$ (k=m)	Entr. Loss $k^* - m$	Provable?
Computat. Secure	SHA/RO	$\epsilon \cdot 2^{m-k}$	$2\epsilon$	0	no
Unpredict.	log(1/ε)- wise hash	$\epsilon \cdot \log(1/\epsilon) \cdot 2^{m-k}$	$\epsilon \cdot \log(1/\epsilon)$	$\log \log(1/\epsilon)$	yes
Square- Friendly	universal hash	$\sqrt{\epsilon} \cdot 2^{m-k}$	$\sqrt{\epsilon}$	$\log(1/\epsilon)$	yes
Computat. Secure	PRG + pairwise hash	$\sqrt{\epsilon_{\text{PRG}}} \cdot 2^{m-k}$	$\epsilon + \sqrt{\epsilon_{\text{PRG}}}$	$\log(\epsilon_{\text{PRG}}/\epsilon^2)$	yes*
ANY	universal hash	$\sqrt{2^{m-k}}$	1	$2 \log(1/\epsilon)$	yes

\* Under standard and minimal cryptographic assumptions (OWFs)



# 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** KDFs require **OWFs** ☹️
- **Main challenge**: better computational KDFs to close theory-vs-practice gap even further

# Questions?





# One-Time Pad

Plaintext  
+ Keyword

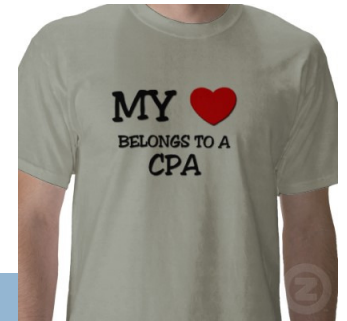
-----  
Ciphertext

44

- **Expect** to fail even for min-entropy  $m - 1$ 
  - $A(c) = c \Rightarrow f(0) = 1/2, f(1) = -1/2 \Rightarrow \epsilon = 0, \sigma = 1/4$
- Similar problem for PRGs/PRFs/PRPs ☹️

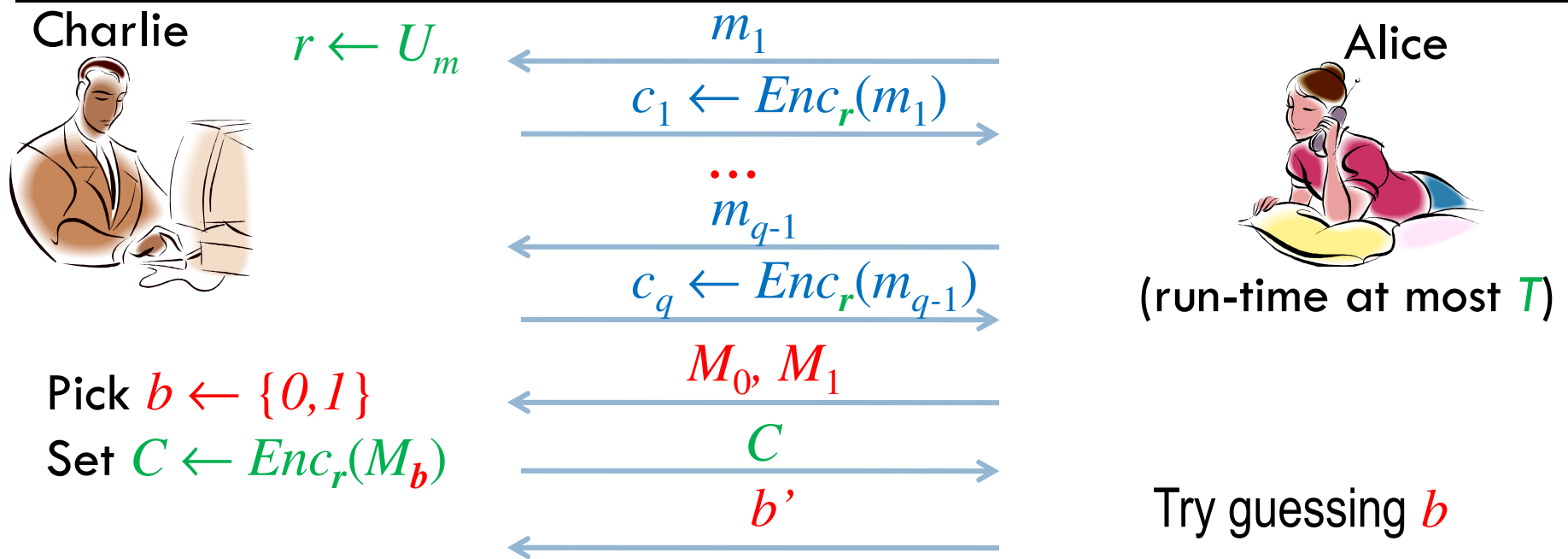


# CPA Security of Encryption



45

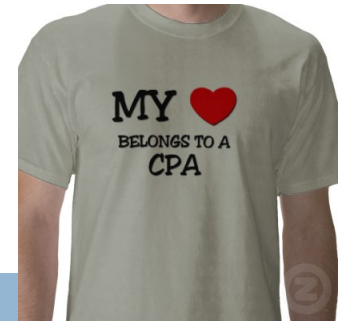
□ Probabilistic Enc/Dec:  $c \leftarrow Enc_r(m) ; m = Dec_r(c)$



□ Define  $f(r) = Adv(A, r) = \Pr[b = b'] - \frac{1}{2} \in [-\frac{1}{2}, \frac{1}{2}]$

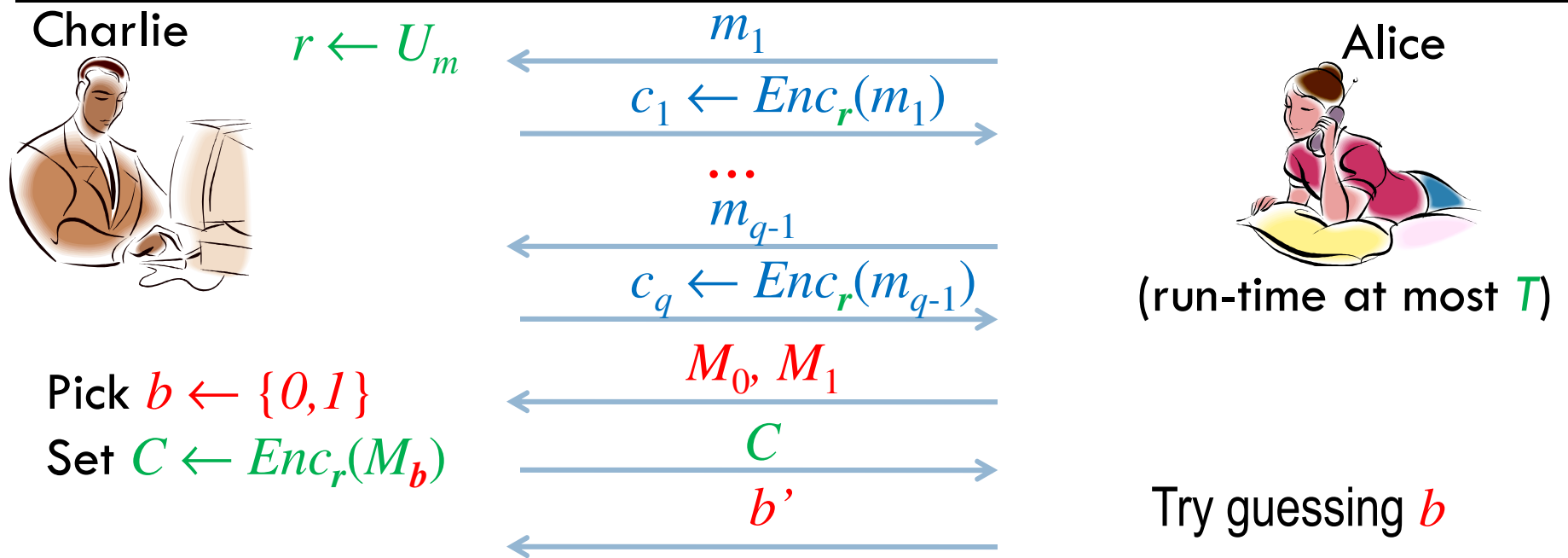
□ Leads to  $(T, q, \epsilon)$ -security /  $(T, q, \sigma)$ -square security

# CPA Security of Encryption



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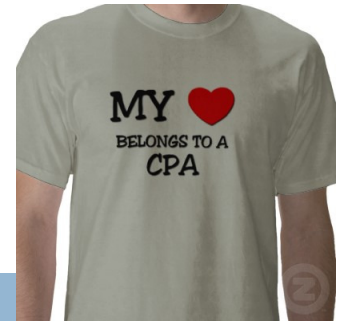
□ Probabilistic Enc/Dec:  $c \leftarrow Enc_r(m) ; m = Dec_r(c)$



□ **Lemma:** if  $Enc$  is  $(2T, 2q, 2\varepsilon)$ -secure, then

$Enc$  is  $(T, q, \varepsilon)$ -square secure (“ $\sigma \approx \varepsilon$ ”)

# Square Security of CPA



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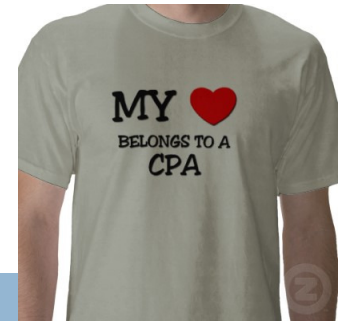
- **Insight:** for any  $A$  making  $q$  encryption queries, there exists  $B$  making  $2q$  encryption queried s.t.

$$\forall r \quad Adv(B, r) = 2Adv(A, r)^2 \geq 0 \quad (**)$$

- Here's  $B$ :

1. Run  $A$  once against *simulated* challenger  $C$ 
  - Choose selection bit **yourself**  $\Rightarrow$  can check if  $A$  "won"
  - Spend  $q$  queries to simulate *both*  $A$  and  $C$
2. Run  $A$  again against *real* challenger  $C$  (+  $q$  queries)
3. If  $A$  lost in Step 1., reverse  $A$ 's guess in Step 2.
  - Intuition: Step 3. ensures  $B$  has advantage  $\geq 0$

# Square Security of CPA



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- **Insight:** for any  $A$  making  $q$  encryption queries, there exists  $B$  making  $2q$  encryption queried s.t.  
 $\forall r \quad Adv(B, r) = 2Adv(A, r)^2 \geq 0 \quad (**)$

- Here's  $B$ :

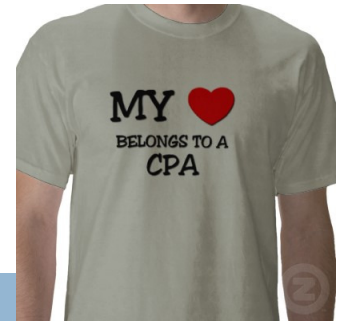
1. Run  $A$  once against *simulated* challenger  $C$
2. Run  $A$  again against *real* challenger  $C$
3. If  $A$  lost in Step 1., reverse  $A$ 's guess in Step 2.

- $\Pr[B \text{ wins}] = \Pr[A \text{ wins twice}] + \Pr[A \text{ looses twice}]$   
$$= \left(\frac{1}{2} \pm \varepsilon\right)^2 + \left(\frac{1}{2} \mp \varepsilon\right)^2 = \frac{1}{2} + 2\varepsilon^2$$





# Square Security of CPA



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- **Insight:** for any  $A$  making  $q$  encryption queries, there exists  $B$  making  $2q$  encryption queries s.t.  
 $\forall r \quad Adv(B, r) = 2Adv(A, r)^2 \geq 0 \quad (**)$

- Hence,  $\sigma = \mathbb{E}[Adv(A, r)^2] \leq \frac{1}{2} \mathbb{E}[Adv(B, r)] \leq \epsilon$

- **Corollary:** if  $Enc$  is  $(2T, 2q, 2\epsilon)$ -secure, then  $Enc$  is  $(T, q, \sqrt{\epsilon \cdot 2^d})$ -secure in the  $(m-d)$ -real model

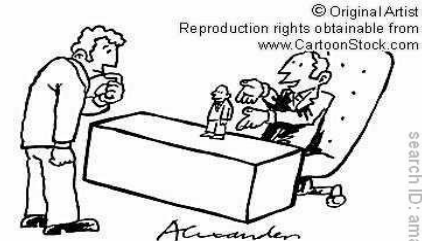
- [BG09]:  $((1 + c^4)T, (1 + c^4)q, \epsilon) \Rightarrow (T, q, O(\frac{1}{c} \cdot \sqrt{\epsilon \cdot 2^d}))$

- Same argument works for weak PRFs, greatly simplifying [Pie09]



# New Dense Model Theorem

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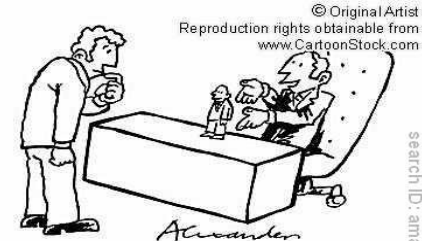
I had my people and your people  
crushed together to create this  
one superdense person

- How to build PRG with weak seed?
  - ▣ Naive:  $G(X)$  **not** pseudorandom, even if  $H_2(X) = m - 1$
- **Dense Model Theorem**: if  $H_\infty(X) \geq m - d$ , then  $G(X)$  has “pseudo-entropy”  $2m - d \gg m$ 
  - ▣ Implies  $G(\text{Ext}(G(X); S))$  is pseudorandom given  $S$
  - ▣ Problem: bad degradation in run-time  $t$
- **Our Version**: if  $H_2(X) \geq m - d$ , then  $G(\text{PIH}_{G(X)}(S))$  is pseudorandom given  $S$ 
  - ▣ No degradation in  $t$ , security  $\sqrt{\epsilon \cdot 2^d}$  (vs.  $\epsilon \cdot 2^d$ )

Pairwise  
independent hash

wprf

# New Dense Model Theorem



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I had my people and your people  
crushed together to create this

Leads to **same** concrete instantiation:  
pairwise independent hash is a good extractor!

**Open:** a **single** unified proof, giving a smooth  
transition between these two “extreme” bounds

- ▣ Implies  $G(\text{Ext}(G(X); S))$  is pseudorandom given  $S$
- ▣ Problem: bad degradation in run-time  $t$
- ▣ **Our Version:** if  $H_2(X) \geq m - d$ , then  $G(\text{PIH}_{G(X)}(S))$   
is pseudorandom given  $S$
- ▣ No degradation in  $t$ , security  $\sqrt{\epsilon \cdot 2^d}$  (vs.  $\epsilon \cdot 2^d$ )

Pairwise  
independent hash

