

Perfectly Secure Encryption of Individual Sequences

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Background and Motivation

The **individual–sequence approach** (with FSM's) to IT has been studied in:

- Data compression (Ziv & Lempel '78,...).
- Source/channel simulation (Martín *et al.* '10, Seroussi '06).
- Classification (Ziv & Merhav, '93).
- Prediction (Feder, Merhav & Gutman '92, ...).
- Denoising (Weissman *et al.*, '05,...).
- Channel coding (Lomnitz & Feder '10, Shayevitz & Feder '05).

Information–theoretic security has been studied **almost** exclusively from the **probabilistic approach**.

Background and Motivation (Cont'd)

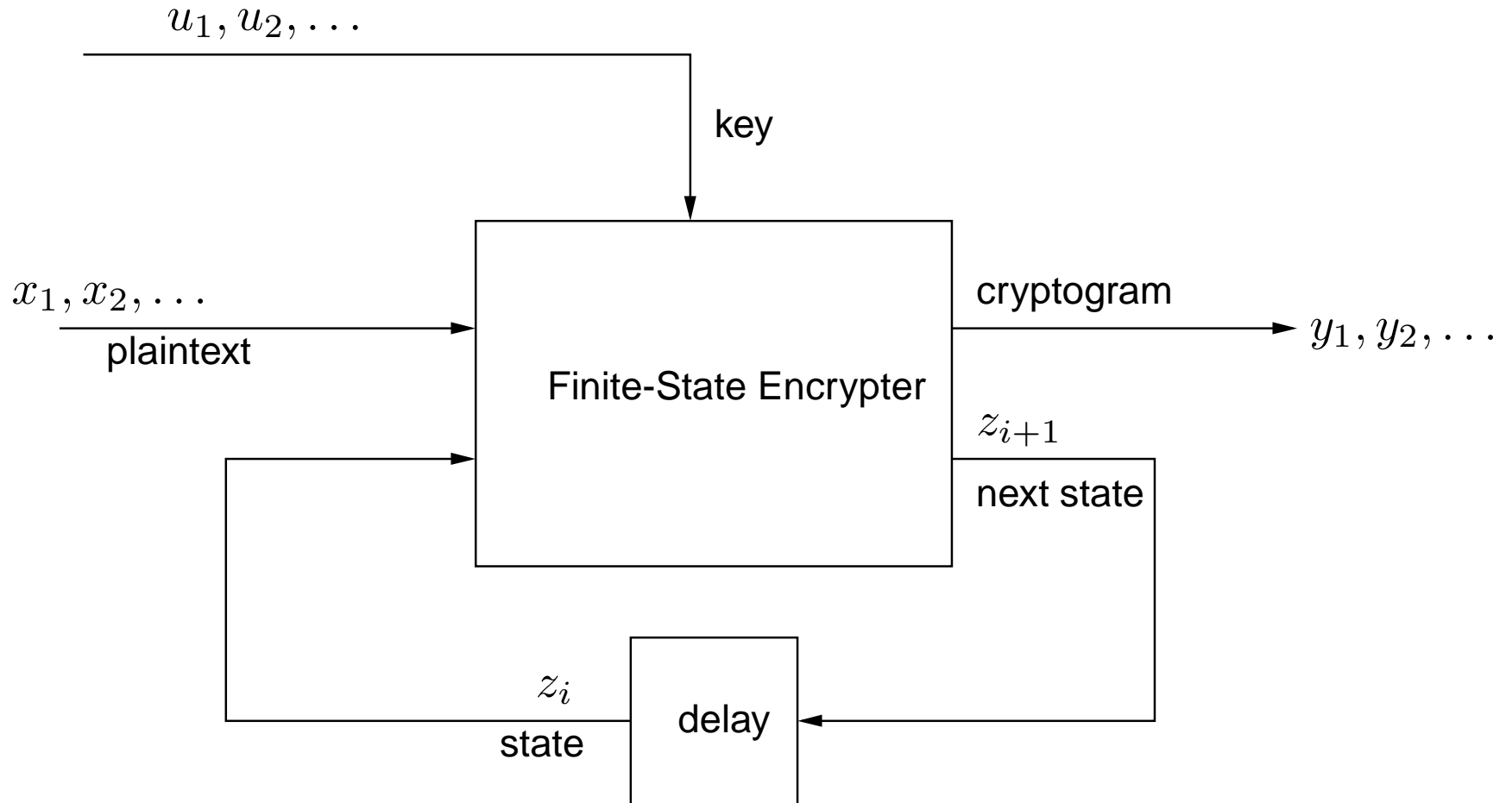
The only exception (a.f.a.i.k.) is an unpublished memorandum by Ziv [1978]:

- Plaintext source – individual sequence.
- Encrypter – general block encoder.
- Prior knowledge: plaintext \rightarrow FSM \rightarrow all-zero sequence.
- Full security: uncertainty – essentially not reduced by cryptogram.
- Main result: minimum needed key rate \sim LZ compressibility.

Encrypter/decrypter have **unlimited resources**, whereas eavesdropper is **limited by FSM**.

Our approach: the other way around – encryption using FSM's.

Finite-State Encrypter Model



$$\begin{aligned}t_i &= t_{i-1} + \Delta(z_i, x_i), & t_0 &\triangleq 0 \\k_i &= (u_{t_{i-1}+1}, u_{t_{i-1}+2}, \dots, u_{t_i}) \\y_i &= f(z_i, x_i, k_i) \\z_{i+1} &= g(z_i, x_i)\end{aligned}$$

Finite–State Encrypter Model (Cont'd)

Perfect security: $\Pr\{y^n | x\}$ – independent of x .

Information losslessness (IL): \exists large n s.t. (z_1, z_{n+1}, k^n, y^n) determines x^n .

Key rate of encrypter E :

$$\sigma_E(x^n) = \frac{1}{n} \sum_{i=1}^n \Delta(z_i, x_i) = \frac{1}{n} \sum_{i=1}^n \ell(k_i).$$

$$\sigma_s(x^n) = \min_{E \in \mathcal{E}(s)} \sigma_E(x^n),$$

where $\mathcal{E}(s)$ = set of all perfectly secure, IL encrypters with $\leq s$ states.

$$\sigma_s(\mathbf{x}) = \limsup_{n \rightarrow \infty} \sigma_s(x^n)$$

Finite–state encryptability: $\sigma(\mathbf{x}) = \lim_{s \rightarrow \infty} \sigma_s(\mathbf{x})$.

Main Result

Let $\rho_{LZ}(x^n)$ denote the LZ compression ratio, i.e.,

$$\rho_{LZ}(x^n) = \frac{c(x^n) \log c(x^n)}{n},$$

where $c(x^n)$ = number of LZ phrases in x^n .

Theorem (converse): For every x^n

$$\sigma_s(x^n) \geq \rho_{LZ}(x^n) - O\left(s \cdot \sqrt{\frac{\log(\log n)}{\log n}}\right).$$

Consequently,

$$\sigma(\mathbf{x}) \geq \rho(\mathbf{x}).$$

Discussion

- Direct: LZ compression + one-time pad encryption – $\sigma(x) = \rho(x)$.
- Natural individual–sequence counterpart to the known probabilistic result.
- Same conclusion as in [Ziv78], although the model is different.
- Upperbound – lowerbound = $O(\sqrt{\log(\log n)/\log n})$.
- In compression – $O(1/\log n)$.

Main Ideas of Proof

Define joint empirical distribution of m -blocks of (x^n, k^n, y^n, \dots) .

$$\sigma_E(x^n) = \frac{\ell(k^n)}{n} = \frac{H(K^m|L)}{m} \geq \frac{1}{m} [H(K^m) - \alpha s \log(m+1)].$$

Using usual information-theoretic arguments (+ IL + full security):

$$H(K^m) \geq H(X^m) - H(Z, Z' | Y^m, K^m) \geq H(X^m) - 2 \log s.$$

Now, since Shannon code = FS encoder:

$$\frac{H(X^m)}{m} \geq \rho_{LZ}(x^n) - \delta_s(m, n).$$

So eventually,

$$\sigma_E(x^n) \geq \rho_{LZ}(x^n) - \text{vanishing terms.}$$

Extensions

Availability of Side Information

Assume that everybody has access to SI s_1, s_2, \dots (individual sequence).
Modifying the model definition:

$$\begin{aligned}t_i &= t_{i-1} + \Delta(z_i, x_i, s_i), & t_0 &\triangleq 0 \\k_i &= (u_{t_{i-1}+1}, u_{t_{i-1}+2}, \dots, u_{t_i}) \\y_i &= f(z_i, x_i, k_i, s_i) \\z_{i+1} &= g(z_i, x_i, s_i)\end{aligned}$$

Perfect security: $\Pr(y^n | x, s)$ – independent of x .

Info losslessness: For large enough n : $(z_1, z_{n+1}, s^n, y^n, k^n)$ determine x^n .

Main result: Same but with $\rho_{LZ}(x^n)$ replaced by $\rho_{LZ}(x^n | s^n)$ – **conditional LZ parsing** of x^n given s^n [Ziv '85].

Achievable even if encrypter does not see s^n : S–W coding + one–time pad.

Availability of SI (Cont'd) – Conditional Parsing

- Apply LZ to $((x_1, s_1), (x_2, s_2), \dots, (x_n, s_n))$; $c(x^n, s^n) =$ number of phrases.
- $c(s^n) =$ number of distinct phrases of s^n .
- $s(l) =$ the l th distinct s -phrase, $l = 1, 2, \dots, c(s^n)$.
- $c_l(x^n | s^n) =$ number of $s(l)$ in parsing of s^n .

$$\rho_{LZ}(x^n | s^n) = \frac{1}{n} \sum_{l=1}^{c(s^n)} c_l(x^n | s^n) \log c_l(x^n | s^n).$$

For example,

$$\begin{aligned} x^6 &= 0 | 1 | 00 | 01 | \\ s^6 &= 0 | 1 | 01 | 01 | \end{aligned}$$

then

$$c(x^6, s^6) = 4, \quad c(s^6) = 3, \quad s(1) = 0, \quad s(2) = 1, \quad s(3) = 01,$$

$$c_1(x^6 | s^6) = c_2(x^6 | s^6) = 1, \quad c_3(x^6 | s^6) = 2.$$

Lossy Reconstruction

Modifications in model definition:

- Legitimate reconstruction \hat{x}^n must satisfy $d(x^n, \hat{x}^n) \leq nD$ w.p. 1.
- Distortion measure – completely arbitrary (need not be even additive).
- IL property can be relaxed to a weaker requirement (details in the paper).
- Perfect security: y^n is statistically independent of both x and \hat{x} .

Main theorem essentially as before but $\rho_{LZ}(x^n)$ should be replaced by

$$r_{LZ}(D; x^n) = \min_{d(x^n, \hat{x}^n) \leq nD} \rho_{LZ}(\hat{x}^n).$$

- Not obvious that best \hat{x}^n is deterministic (could have depended on key).
- Achievability: again, conceptually obvious.
- In the full paper: also SI + lossy reconstruction; No longer based on LZ..

Thank You!