

DD2448 Foundations of Cryptography

Lecture 3

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Linear Cryptanalysis of the SPN

Basic Idea – Linearize

Find an expression of the following form with a high probability of occurrence.

$$P_{i_1} \oplus \cdots \oplus P_{i_p} \oplus C_{j_1} \oplus \cdots \oplus C_{j_c} = K_{\ell_1, s_1} \oplus \cdots \oplus K_{\ell_k, s_k}$$

Each random plaintext/ciphertext pair gives an estimate of

$$K_{\ell_1, s_1} \oplus \cdots \oplus K_{\ell_k, s_k}$$

Collect many pairs and make a better estimate based on the majority vote.

How do we come up with the desired expression?

How do we compute the required number of samples?

Definition. The bias $\epsilon(X)$ of a binary random variable X is defined by

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$\approx 1/\epsilon^2(X)$ samples are required to estimate X
(Matsui)

Linear Approximation of S-Box (1/3)

Let X and Y be the input and output of an S-box, i.e.

$$Y = S(X) .$$

We consider the bias of linear combinations of the form

$$a \cdot X \oplus b \cdot Y = \left(\bigoplus_i a_i X_i \right) \oplus \left(\bigoplus_i b_i Y_i \right) .$$

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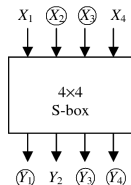
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Example: $X_2 \oplus X_3 = Y_1 \oplus Y_3 \oplus Y_4$

The expression holds in 12 out of the 16 cases. Hence, it has a bias of $(12 - 8)/16 = 4/16 = 1/4$.



Linear Approximation of S-Box (2/3)

- ▶ Let $N_L(a, b)$ be the number of zero-outcomes of $a \cdot X \oplus b \cdot Y$.
- ▶ The bias is then

$$\epsilon(a \cdot X \oplus b \cdot Y) = \frac{N_L(a, b) - 8}{16},$$

since there are four bits in X , and Y is determined by X .

Linear Approximation Table (3/3)

$$N_L(a, b) - 8$$

		Output Sum															
		0	1	2	3	4	5	6	7	8	9	A	B	C	D	E	F
I n p u t	0	+8	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
	1	0	0	-2	-2	0	0	-2	+6	+2	+2	0	0	+2	+2	0	0
	2	0	0	-2	-2	0	0	-2	-2	0	0	+2	+2	0	0	-6	+2
	3	0	0	0	0	0	0	0	0	+2	-6	-2	-2	+2	+2	-2	-2
	4	0	+2	0	-2	-2	-4	-2	0	0	-2	0	+2	+2	-4	+2	0
	5	0	-2	-2	0	-2	0	+4	+2	-2	0	-4	+2	0	-2	-2	0
	6	0	+2	-2	+4	+2	0	0	+2	0	-2	+2	+4	-2	0	0	-2
	7	0	-2	0	+2	+2	-4	+2	0	-2	0	+2	0	+4	+2	0	+2
	8	0	0	0	0	0	0	0	0	-2	+2	+2	-2	+2	-2	-2	-6
	9	0	0	-2	-2	0	0	-2	-2	-4	0	-2	+2	0	+4	+2	-2
	A	0	+4	-2	+2	-4	0	+2	-2	+2	+2	0	0	+2	+2	0	0
	B	0	+4	0	-4	+4	0	+4	0	0	0	0	0	0	0	0	0
	C	0	-2	+4	-2	-2	0	+2	0	+2	0	+2	+4	0	+2	0	-2
	D	0	+2	+2	0	-2	+4	0	+2	-4	-2	+2	0	+2	0	0	+2
	E	0	+2	+2	0	-2	-4	0	+2	-2	0	0	-2	-4	+2	-2	0
	F	0	-2	-4	-2	-2	0	+2	0	0	-2	+4	-2	-2	0	+2	0

This gives linear approximation for one round.

How do we come up with linear approximation for more rounds?

Piling-Up Lemma

Lemma. Let X_1, \dots, X_t be independent binary random variables and let $\epsilon_i = \epsilon(X_i)$. Then

$$\epsilon \left(\bigoplus_i X_i \right) = 2^{t-1} \prod_i \epsilon_i .$$

Proof. Case $t = 2$:

$$\begin{aligned} \Pr[X_1 \oplus X_2 = 0] &= \Pr[(X_1 = 0 \wedge X_2 = 0) \vee (X_1 = 1 \wedge X_2 = 1)] \\ &= \left(\frac{1}{2} + \epsilon_1\right)\left(\frac{1}{2} + \epsilon_2\right) + \left(\frac{1}{2} - \epsilon_1\right)\left(\frac{1}{2} - \epsilon_2\right) \\ &= \frac{1}{2} + 2\epsilon_1\epsilon_2 . \end{aligned}$$

By induction $\Pr[X_1 \oplus \dots \oplus X_t = 0] = \frac{1}{2} + 2^{t-1} \prod_i \epsilon_i$

Four linear approximations with $|\epsilon_i| = 1/4$

$$S_{12} : X_1 \oplus X_3 \oplus X_4 = Y_2$$

$$S_{22} : X_2 = Y_2 \oplus Y_4$$

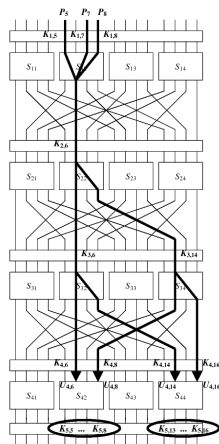
$$S_{32} : X_2 = Y_2 \oplus Y_4$$

$$S_{34} : X_2 = Y_2 \oplus Y_4$$

Combine them to get:

$$U_{4,6} \oplus U_{4,8} \oplus U_{4,14} \oplus U_{4,16} \oplus P_5 \oplus P_7 \oplus P_8 = \bigoplus K_{i,j}$$

with bias $|\epsilon| = 2^{4-1} \left(\frac{1}{4}\right)^4 = 2^{-5}$



- ▶ Our expression (with bias 2^{-5}) links plaintext bits to input bits to the 4th round
- ▶ Partially undo the last round by guessing the last key. Only 2 S-Boxes are involved, i.e., $2^8 = 256$ guesses
- ▶ For a correct guess, the equation holds with bias 2^{-5} . For a wrong guess, it holds with bias zero (i.e., probability close to $1/2$).

Attack Idea

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Required pairs $2^{10} \approx 1000$

Attack complexity $2^{18} \ll 2^{32}$ operations

Linear Cryptanalysis Summary

1. Find linear approximation of S-Boxes.
2. Compute bias of each approximation.
3. Find linear trails.
4. Compute bias of linear trails.
5. Compute data and time complexity.
6. Estimate key bits from many plaintext-ciphertexts pairs.

Linear cryptanalysis is a **known plaintext attack**.

Ideal Block Cipher

Definition. A function $\epsilon(n)$ is negligible if for every constant $c > 0$, there exists a constant n_0 , such that

$$\epsilon(n) < \frac{1}{n^c}$$

for all $n \geq n_0$.

Motivation. Events happening with negligible probability can not be exploited by polynomial time algorithms! (they “never” happen)

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Definition. A family of functions $F : \{0, 1\}^k \times \{0, 1\}^n \rightarrow \{0, 1\}^n$ is pseudo-random if for all polynomial time oracle adversaries A

$$\left| \Pr_K \left[A^{F_K(\cdot)} = 1 \right] - \Pr_{R: \{0,1\}^n \rightarrow \{0,1\}^n} \left[A^{R(\cdot)} = 1 \right] \right|$$

is negligible.

Pseudo-Random Permutation

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$$\left| \Pr_K \left[A^{P_K(\cdot), P_K^{-1}(\cdot)} = 1 \right] - \Pr_{\Pi \in \mathcal{S}_{2n}} \left[A^{\Pi(\cdot), \Pi^{-1}(\cdot)} = 1 \right] \right|$$

is negligible, where \mathcal{S}_{2n} is the set of permutations of $\{0, 1\}^n$.

Definition. Feistel round (H for “Horst Feistel”).

$$H_{F_K}(L, R) = (R, L \oplus F(R, K))$$

Idealized Four-Round Feistel Network

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Theorem. (Luby and Rackoff) If F is a pseudo-random family of functions, then

$$H_{F_{k_1}, F_{k_2}, F_{k_3}, F_{k_4}}(x) = H_{F_{k_4}}(H_{F_{k_3}}(H_{F_{k_2}}(H_{F_{k_1}}(x))))$$

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Why do we need four rounds?