




# CS 553

## CRYPTOGRAPHY

### Lecture 10

More on Linear Cryptanalysis

Instructor  
Dr. Dhiman Saha

- ▶ The idea of linear masks
- ▶ The notion of approximation
- ▶ Expressing key bits in terms of plaintext and ciphertexts
- ▶ Approximating the non-linear component 
- ▶ Extending the approximation to other associated parts of a simple cryptosystem
- ▶ Using the linear approximation to mount a KPA
- ▶ Recovering a single bit of key material

## Sypher00B

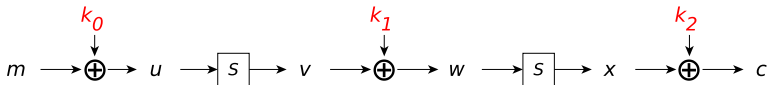
Moving on to a more complex but still toy cryptosystem:

- Sypher00B encrypts 4 bits with **three** 4 bit keys

S-box

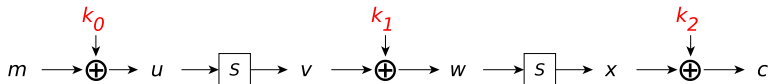
$x$	0	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
$S(x)$	f	e	b	c	6	d	7	8	0	3	9	a	4	2	1	5


## Encryption



- Again, same as Sypher002 with a different SBox

- KPA assumption: attacker knows message  $m$  and ciphertext  $c$



- WLOG the following holds for any mask  $\alpha, \beta, \gamma$ . Why? 

$$(\alpha \cdot m) = (\alpha \cdot k_0) \oplus (\alpha \cdot u) \quad (1)$$

$$(\beta \cdot v) = (\beta \cdot k_1) \oplus (\beta \cdot w) \quad (2)$$

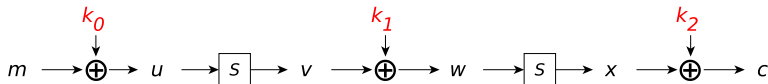
$$(\gamma \cdot x) = (\gamma \cdot k_2) \oplus (\gamma \cdot c) \quad (3)$$


- We assume: we can find  $\alpha, \beta, \gamma$  such that 

$$\alpha \cdot u = \beta \cdot S[u] = \beta \cdot v \quad \text{Holds with prob. } p_1 \neq \frac{1}{2}$$

$$\beta \cdot w = \gamma \cdot S[w] = \gamma \cdot x \quad \text{Holds with prob. } p_2 \neq \frac{1}{2}$$

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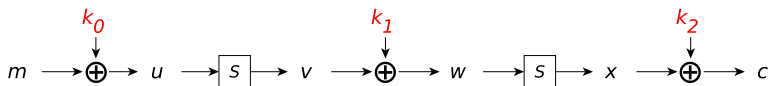
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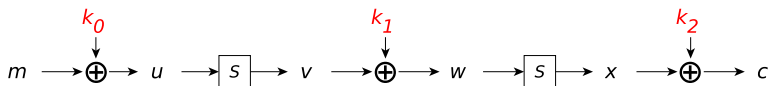
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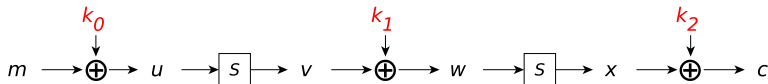
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- Using Eqn. (1) – (3)

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- Note RHS is a constant, for LHS, we know:

$$\left. \begin{array}{ll} (\alpha \cdot u) = (\beta \cdot v) & \text{with prob. } p_1 \\ (\beta \cdot w) = (\gamma \cdot x) & \text{with prob. } p_2 \end{array} \right\} \text{taken over all } u \text{ and } w$$

- Possibility to remove intermediate variables:

$$\alpha \cdot u \quad \beta \cdot v \quad \beta \cdot w \quad \gamma \cdot x$$

What is the probability that the approximation holds when all the internal variables have canceled out?

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Events  $(\alpha \cdot u) = (\beta \cdot v)$  and  $(\beta \cdot w) = (\gamma \cdot x)$  are independent

### The Possibilities

#### Case 1

$$\begin{aligned}(\alpha \cdot u) &= (\beta \cdot v) \\ (\beta \cdot w) &= (\gamma \cdot x)\end{aligned}$$

#### Case 2

$$\begin{aligned}(\alpha \cdot u) &= (\beta \cdot v) \oplus 1 \\ (\beta \cdot w) &= (\gamma \cdot x) \oplus 1\end{aligned}$$

Implication:  $(\alpha \cdot m) \oplus (\gamma \cdot c) = (\alpha \cdot k_0) \oplus (\beta \cdot k_1) \oplus (\gamma \cdot k_2)$

#### Case 3

$$\begin{aligned}(\alpha \cdot u) &= (\beta \cdot v) \\ (\beta \cdot w) &= (\gamma \cdot x) \oplus 1\end{aligned}$$

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     $\begin{cases} \text{Case 1} \rightarrow \text{Prob.} = p_1 \times p_2 \\ \text{Case 2} \rightarrow \text{Prob.} = (1 - p_1) \times (1 - p_2) \end{cases}$
- ▶ Probability of linear approximation:  $p_1 p_2 + (1 - p_1)(1 - p_2)$
- ▶ If  $p_1 = \frac{1}{2} + \epsilon_1$  and  $p_2 = \frac{1}{2} + \epsilon_2$ , then

$$\begin{aligned} & p_1 p_2 + (1 - p_1)(1 - p_2) \\ &= 1 - p_1 - p_2 + 2p_1 p_2 \\ &= 1 - \frac{1}{2} - \epsilon_1 - \frac{1}{2} - \epsilon_2 + 2 \left( \frac{1}{4} + \frac{\epsilon_1}{2} + \frac{\epsilon_2}{2} + \epsilon_1 \epsilon_2 \right) \\ &= \frac{1}{2} + 2\epsilon_1 \epsilon_2 \end{aligned}$$

Intuition

What would the general case look like?




- ▶ Favorable events: 
  - Case 1  $\rightarrow$  Prob. =  $p_1 \times p_2$
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- ▶ Extending this to  $m$  independent events with probabilities  $p_i, i = 1, \dots, m$ , we have:


$$\frac{1}{2} + 2^{m-1} \prod_{i=1}^m \left( p_i - \frac{1}{2} \right) \triangle$$

- ▶ How would you define the event in the general case?
  - ▶ What is actually piling-up? 
  - ▶ What happens when constituent events are true?

The piling-up lemma allows us to compute the **bias** of a set of combined linear approximations provided that the constituent linear approximations are **independent**.

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- ▶ Task is to find masks  $\alpha, \beta, \gamma$  for satisfying

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- ▶ How?

### The Linear Approximation Table

This table lists the probabilities that the **sum** of certain input bits of  $a$  equals the sum of certain output bits of  $S[a]$ .

- ▶ Each entry gives us the linear characteristic for a pair of input-output masks

$$\alpha \xrightarrow{S} \beta$$

- ▶ And also the associated **bias**

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# The Linear Approximation Table

	1	2	3	4	5	6	7	8	9	a	b	c	d	e	f
1	-2	.	2	.	-2	4	-2	2	4	2	.	-2	.	2	.
2	2	-2	.	-2	.	.	2	2	4	.	2	4	-2	-2	.
3	4	2	2	-2	2	.	.	.	.	2	-2	-2	-2	.	4
4	.	-2	2	2	-2	.	.	-4	.	2	2	2	2	.	4
5	-2	2	.	2	4	.	2	-2	4	.	-2	.	2	-2	.
6	-2	.	2	.	2	4	2	2	-4	2	.	2	.	-2	.
7	.	.	.	4	.	-4	.	.	.	.	4	.	4	.	.
8	.	-2	2	-4	.	2	2	-4	.	-2	-2	.	.	2	-2
9	-2	-6	.	.	2	-2	.	2	.	.	-2	-2	.	.	2
a	-2	.	-6	-2	.	2	.	-2	.	2	.	.	-2	.	2
b	.	.	.	2	-2	2	-2	.	.	-4	-4	2	-2	-2	2
c	.	.	.	-2	-2	-2	-2	.	.	4	-4	2	2	-2	-2
d	-2	.	2	2	.	-2	.	-2	.	2	.	.	-6	.	-2
e	2	-2	.	.	2	2	-4	-2	.	.	2	-2	.	-4	-2
f	-4	2	2	-4	.	-2	-2	.	.	-2	2	.	.	-2	2

Highest Bias for  $\alpha = \beta = \gamma = d$

- ▶ The chosen characteristic:  $d \xrightarrow{S} d \xrightarrow{S} d$
- ▶ For Sypher00B this implies:

$$(d \cdot m) \oplus (d \cdot c) = (d \cdot k_0) \oplus (d \cdot k_1) \oplus (d \cdot k_2)$$

- ▶ Associated prob.

$$\begin{aligned} \Pr(d \xrightarrow{S} d \xrightarrow{S} d) &= \frac{1}{8} \times \frac{1}{8} + \frac{7}{8} \times \frac{7}{8} \\ &= \frac{25}{32} \\ &= \frac{1}{2} + \frac{9}{32} \end{aligned}$$

- ▶ Attacker collects  $N$  KPs to calculate  $(d \cdot m) \oplus (d \cdot c)$
- ▶ Based on counter values determine if  $(k_0 \oplus k_1 \oplus k_2) \cdot d \stackrel{?}{=} 0/1$

- ▶ The chosen characteristic:  $d \xrightarrow{S} d \xrightarrow{S} d$
- ▶ For Sypher00B this implies:


$$(d \cdot m) \oplus (d \cdot c) = (d \cdot k_0) \oplus (d \cdot k_1) \oplus (d \cdot k_2)$$

- ▶ Associated prob.

$$\begin{aligned} \Pr \left( d \xrightarrow{S} d \xrightarrow{S} d \right) &= \frac{1}{8} \times \frac{1}{8} + \frac{7}{8} \times \frac{7}{8} \\ &= \frac{25}{32} \\ &= \frac{1}{2} + \frac{9}{32} \end{aligned}$$

- ▶ Attacker collects  $N$  KPs to calculate  $(d \cdot m) \oplus (d \cdot c)$
- ▶ Based on counter values determine if  $(k_0 \oplus k_1 \oplus k_2) \cdot d \stackrel{?}{=} 0/1$


# Is recovering single key-bit material enough?

- ▶ Can we do more? How?
- ▶ How about more linear approximations? 
- ▶ Fine, but may not be always available

Better if we can use one but recover more than one

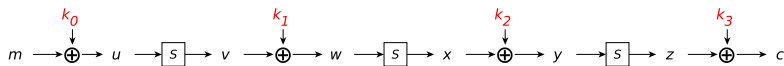
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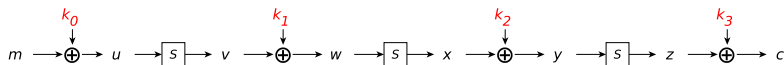
- ▶ Holds with prob.  $\frac{1}{2} + \frac{9}{32}$
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Guess  $k_3$  and invert last round

- ▶ Repeat the attack used for Sypher00B for every guess of  $k_3$





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$$y' = S^{-1}[c \oplus i]$$

for every guess  $k_3 = i$

- Uses the corresponding message  $m$  to compute

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- For each guess  $i$ , he maintains two counters  $T_0^i$  and  $T_1^i$

$$T_0^i \text{++ if } (d \cdot m) \oplus (d \cdot y') = 0$$

$$T_1^i \text{++ if } (d \cdot m) \oplus (d \cdot y') = 1$$

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## How to distinguish?

- ▶ For the correct guess,  $k_3 = \nu$  (say), expected value of

$$\begin{cases} T_0^\nu \leftarrow \frac{N}{2} + \frac{9N}{32} \\ T_1^\nu \leftarrow \frac{N}{2} - \frac{9N}{32} \end{cases}$$

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- ▶ What else?
- ▶ Looking at actual values of  $T_0^i$  and  $T_1^i$  in the highest imbalanced counter value of  $(k_0 \oplus k_1 \oplus k_2) \cdot d$  is recovered
- ▶ For Sypher00C, largest counter indicates value of  $(k_0 \oplus k_1 \oplus k_2) \cdot d$

How many counters? 

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Point to Ponder

Can the same be done for the first round?

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Point to Ponder


Can the **same** be done for the first round?

What is estimate for number of KPs required:  $N$ ?

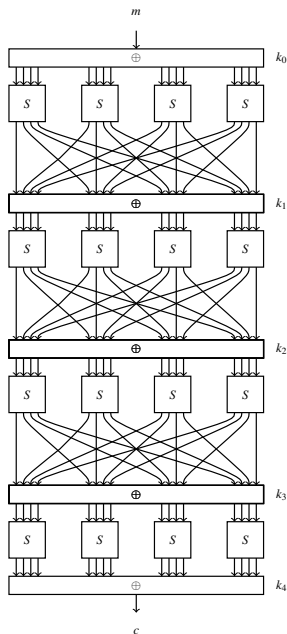
- For Single-bit recovery a good estimate is

$$N = c \left| p - \frac{1}{2} \right|^{-2} \quad \text{or} \quad N = c |\epsilon|^{-2}$$

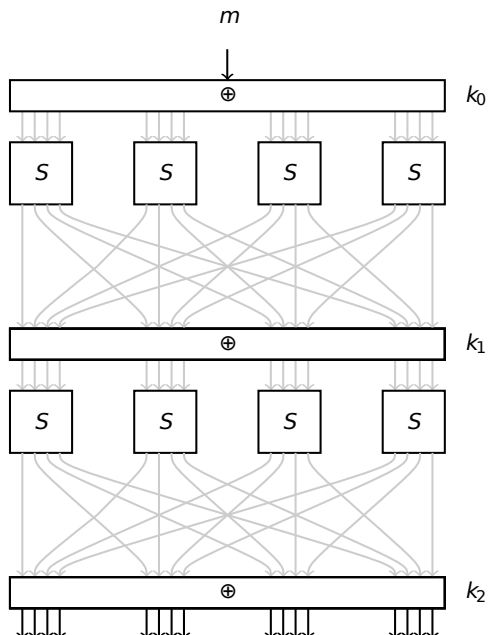
where  $\epsilon \rightarrow$  bias

- Constant  $c \geq 2$  varies with block cipher and attack
- $c$  for single-bit recovery will definitely be less than that for multiple-bit recovery 
- Think about #counters to choose from



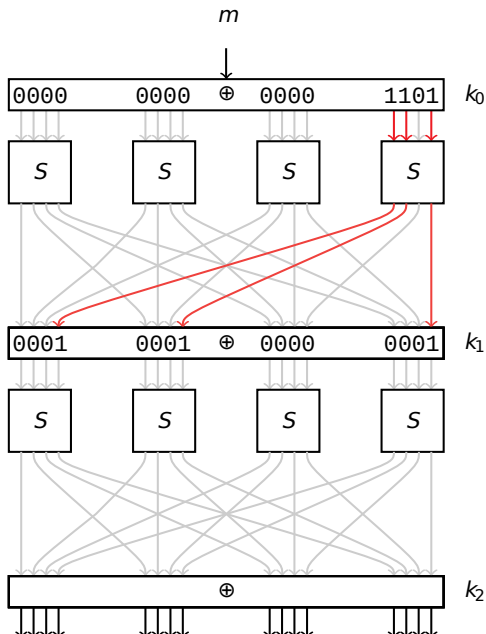


- ▶ Sbox same as Sypher00A-C
- ▶ **Permutation** same as **Sypher004** in DC lecture
- ▶ Number of rounds is 4



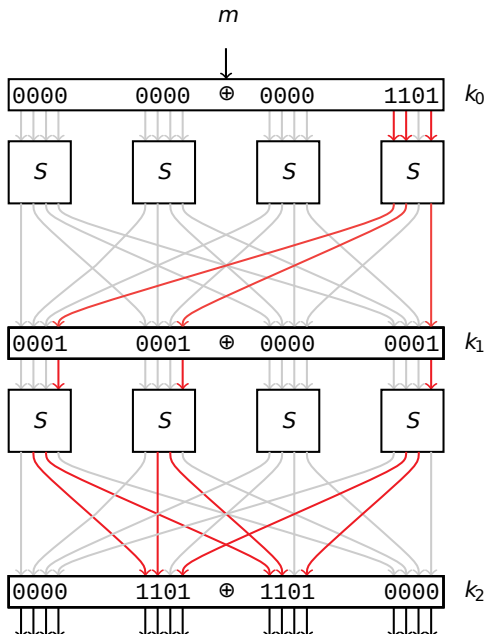
$$(0, 0, 0, d) \xrightarrow{\mathcal{R}} (1, 1, 0, 1)$$

$$p_1 = \frac{1}{2} - \frac{6}{16}$$

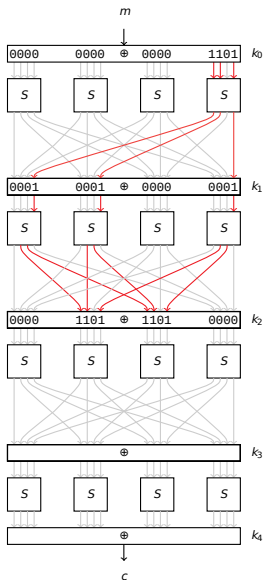


$$(1, 1, 0, 1) \xrightarrow{\mathcal{R}} (0, d, d, 0)$$

$$p_2 = \frac{1}{2} + 2^2 \left( \frac{4}{16} \right)^3 = \frac{1}{2} + \frac{1}{16}$$



# 2-Round Linear Characteristic



$$\blacktriangleright (0, 0, 0, d) \xrightarrow{\mathcal{R}} (1, 1, 0, 1) \xrightarrow{\mathcal{R}} (0, d, d, 0)$$

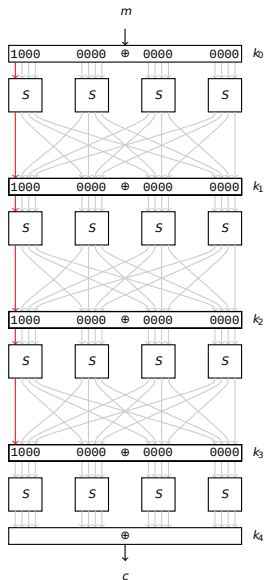
$$p_1 = \frac{1}{2} - \frac{6}{16} = \frac{1}{8}$$

$$p_2 = \frac{1}{2} + 2^2 \left( \frac{4}{16} \right)^3 = \frac{1}{2} + \frac{1}{16} = \frac{9}{16}$$

$\blacktriangleright$  Prob. of 2-round characteristic:

$$\frac{1}{8} \times \frac{9}{16} + \frac{7}{8} \times \frac{7}{16} = \frac{29}{64} = \frac{1}{2} - \frac{3}{64}$$

$\blacktriangleright$  And so on.



- Prob. of one round characteristic:

$$(8, 0, 0, 0) \xrightarrow{\mathcal{R}} (8, 0, 0, 0) : \frac{1}{2} - \frac{4}{16}$$

### Iterative Characteristic

Input mask = Output mask

- Using piling-up lemma we have  
prob. for  $(8, 0, 0, 0) \xrightarrow{3\mathcal{R}} (8, 0, 0, 0)$

$$\frac{1}{2} + 2^2 \left( \frac{1}{4} \right)^3 = \frac{9}{16} = \frac{1}{2} + \frac{1}{16}$$

- Key recovery as illustrated before