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Improving Differential-Neural Cryptanalysis for Large-State SPECK

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1. Preliminaries

- SPECK
- Gohr's Differential-Neural Cryptanalysis
- Differential-Neural Cryptanalysis for Large-State Block Cipher

2. A Parallelizable Key Recovery Framework for Large-State SPECK

- New Training Strategy for Neural Distinguishers
- Parallelizable Key Recovery Framework

3. Improved Partial Neural Distinguisher for Ciphertext

4. Summary

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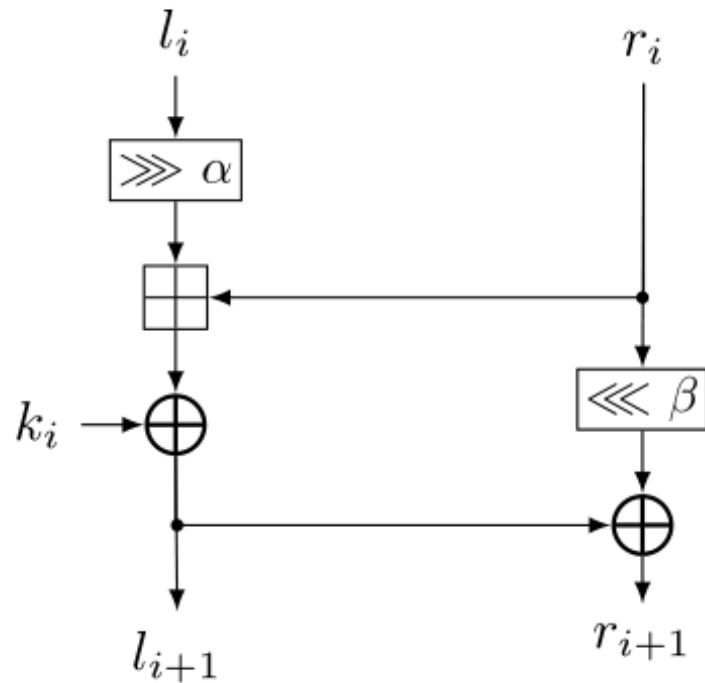
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SPECK is a set of lightweight block cipher algorithms designed by the National Security Agency (NSA) of the United States.



Based on **the block size and key size**, SPECK can be categorized into various versions, such as **SPECK32/64**.

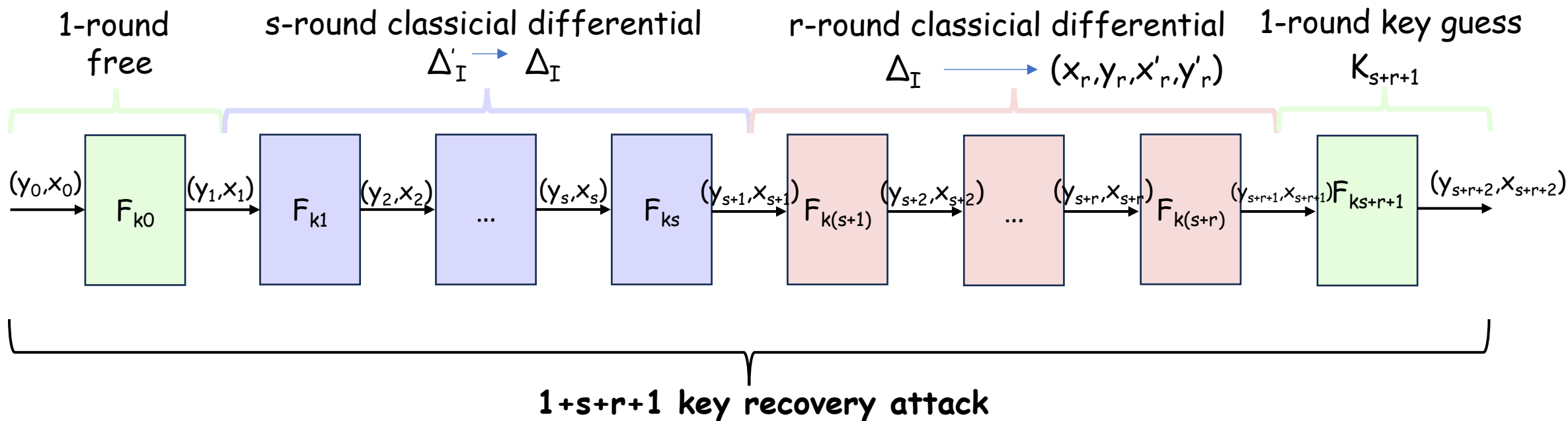
➤ Differential-Neural Distinguishers [Gohr in 2019]:

Task: distinguishing two types of ciphertext pairs

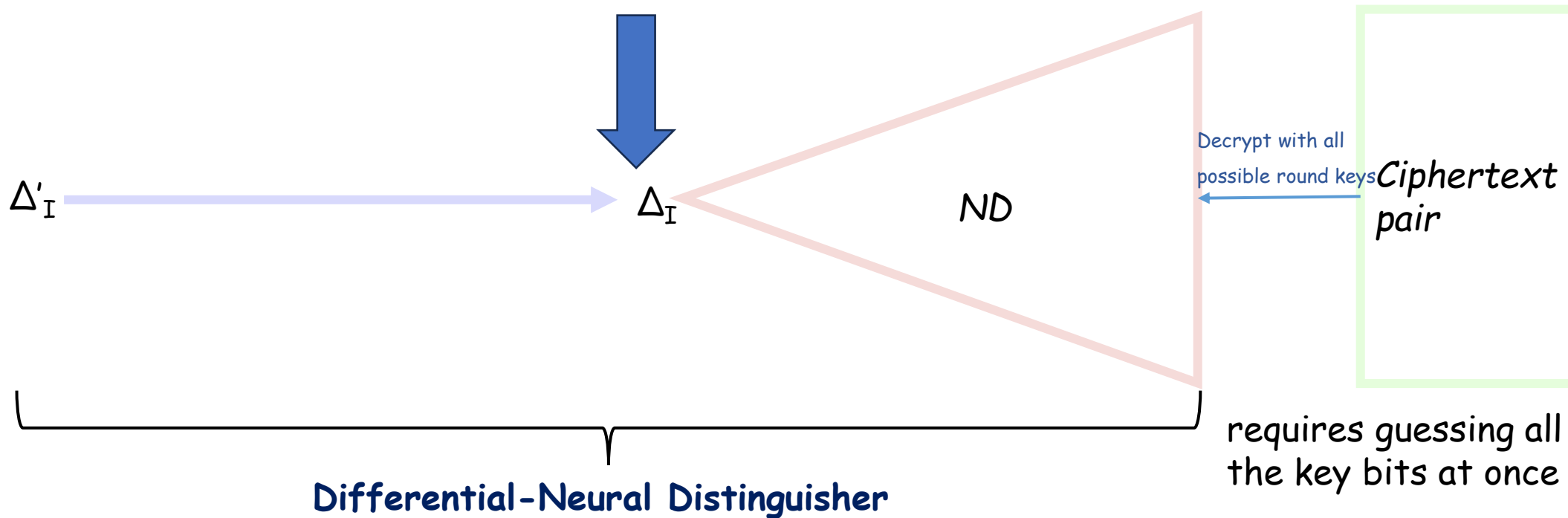
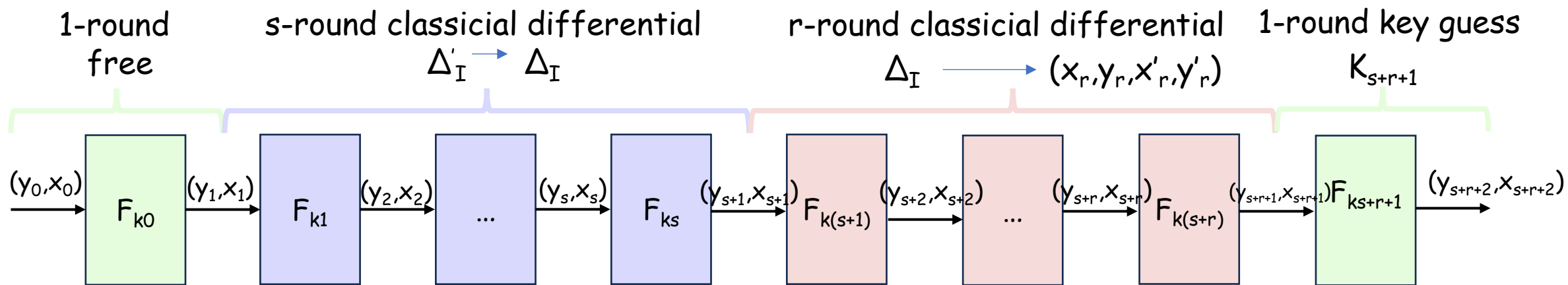
Positive $(C, C'), Y=1$, where $(C, C') \xleftarrow{\text{Enc}} ((P, P') \mid P = \$, P' = P \oplus \Delta_I)$

Negative $(C, C'), Y=0$, where $(C, C') \xleftarrow{\text{Enc}} ((P, P') \mid P = \$, P' = \$')$

➤ Key Recovery Attack [Gohr in 2019] :



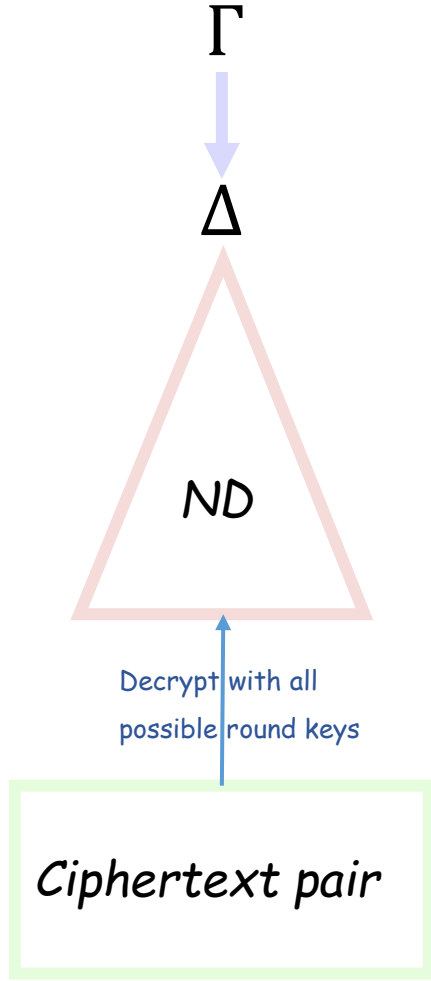
Gohr's Differential-Neural Cryptanalysis



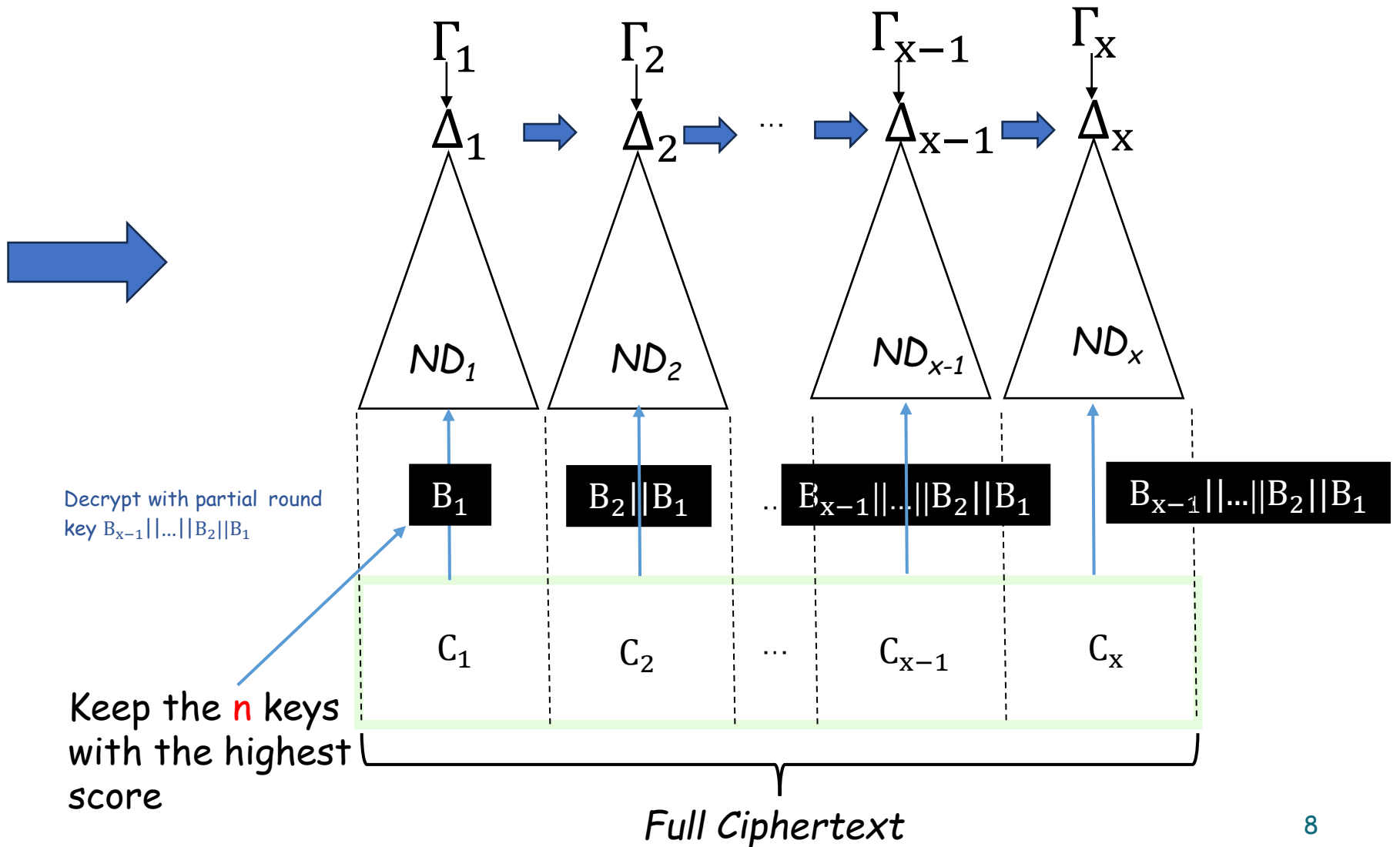
Cryptanalysis for Large-State Block Cipher



Differential-Neural Distinguisher for SPECK32/64



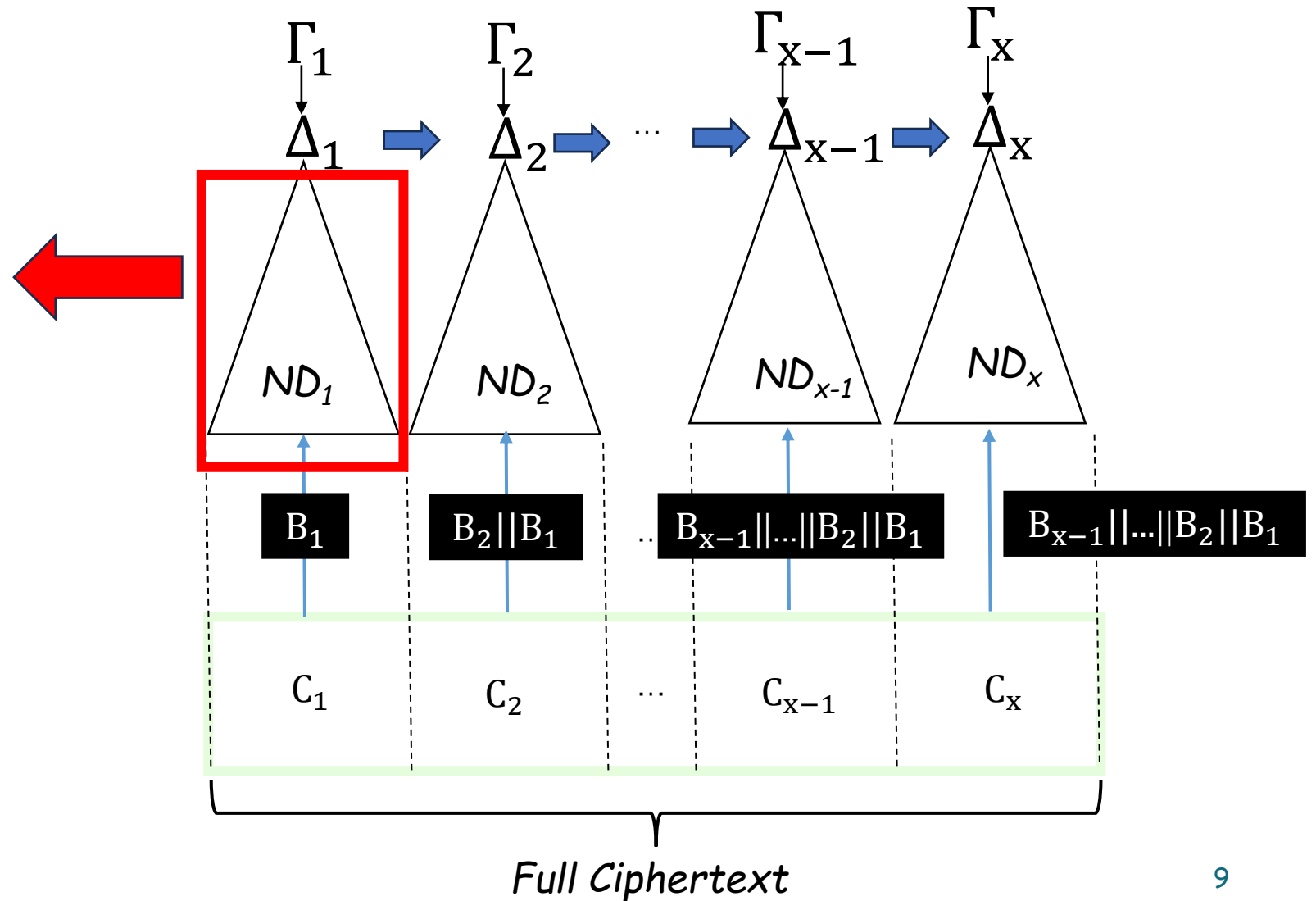
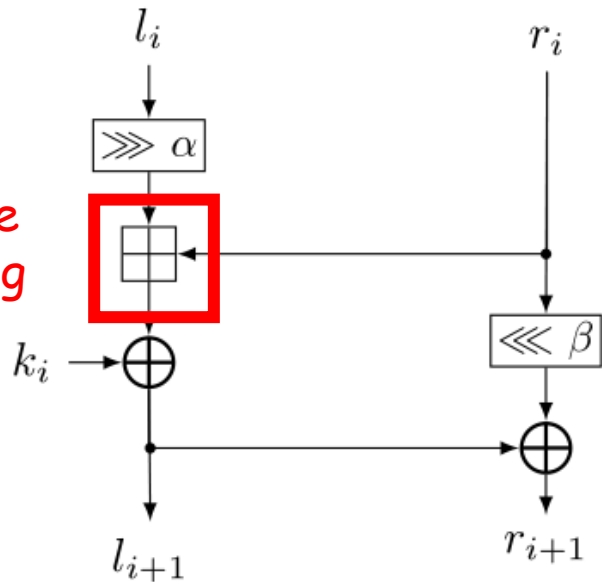
Key Recovery Framework for Large-State SPECK [Chen et al. in 2019]:



Chen et al.'s Key Recovery Framework for Large-State SPECK

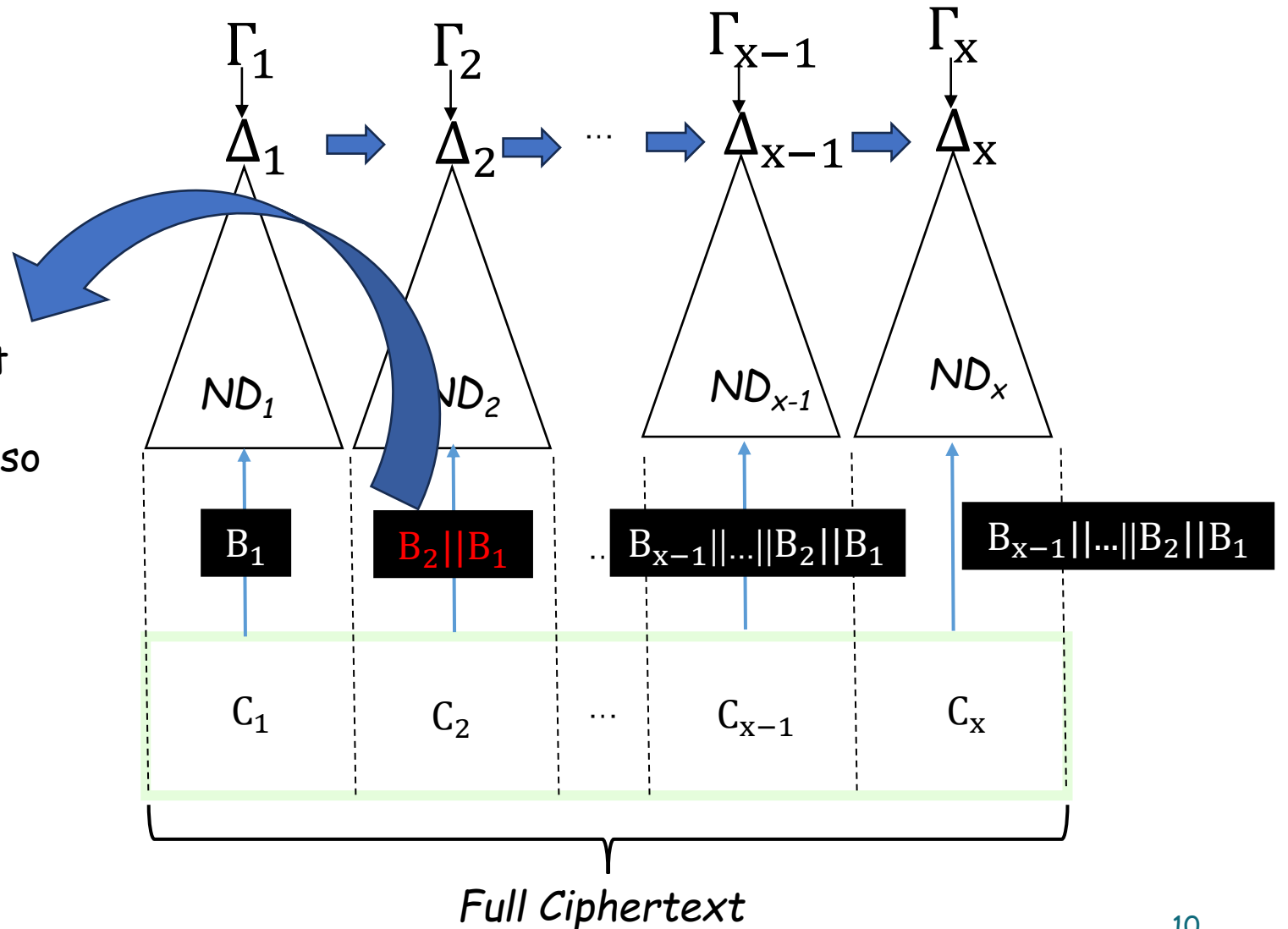
During training, the C_i part of the full ciphertext is used directly for training.

Produce carrying



Chen et al.'s Key Recovery Framework for Large-State SPECK

When recovering the high-level part of the key such as B_2 , not only B_2 should be used to decrypt C_2 , but also the carry effect of using B_1 to decrypt C_1 should be considered.



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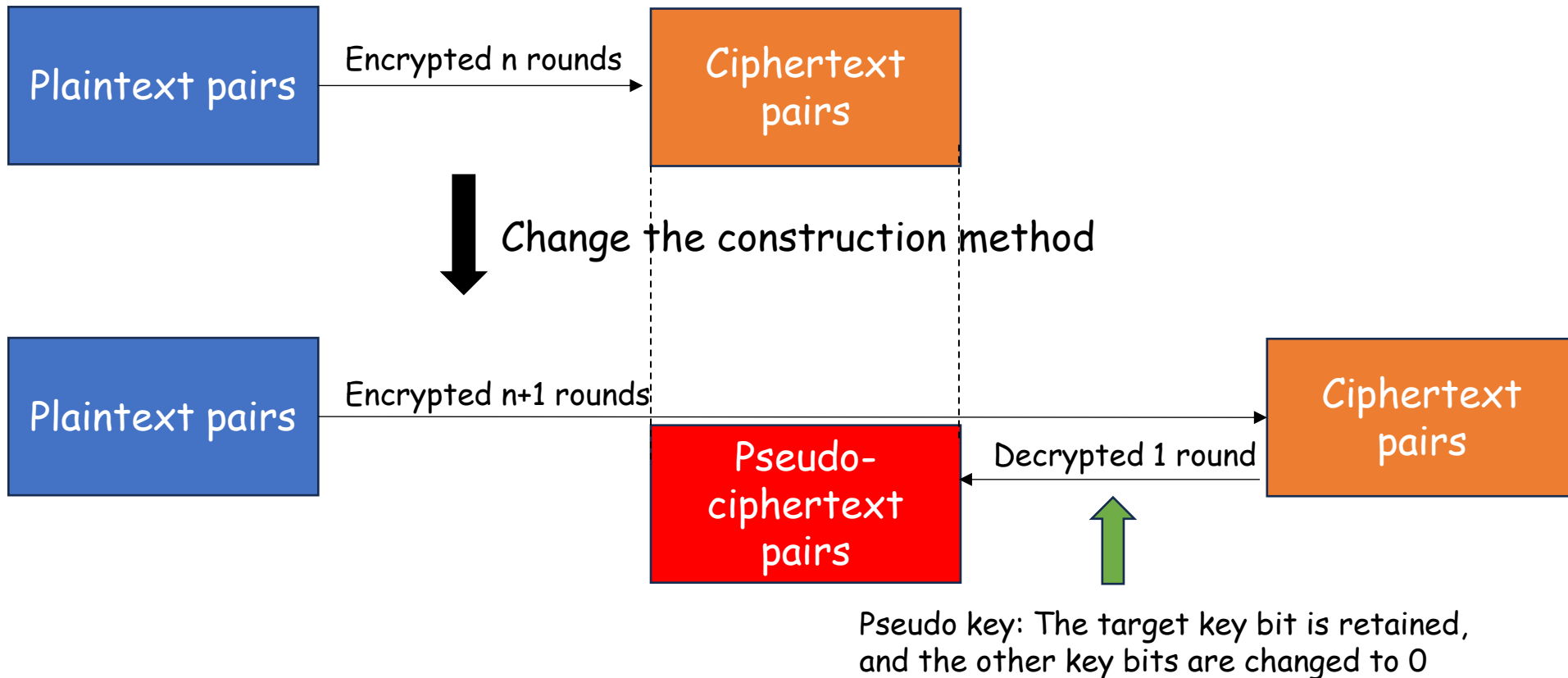
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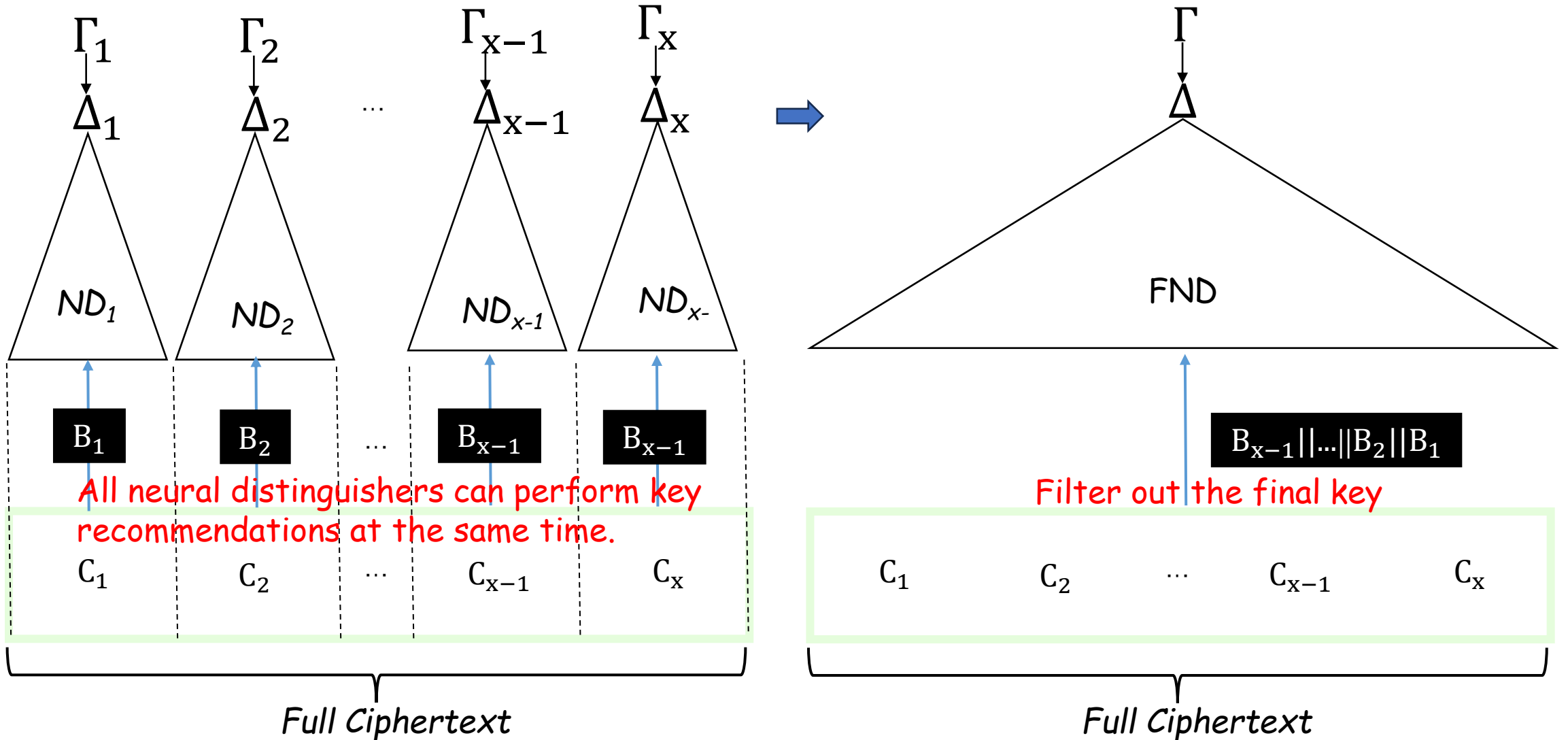
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➤ Change the training data of the partial neural distinguisher



When the neural distinguisher trained by this kind of data does key recovery, it only needs to pay attention to the target key bit, and the other positions are set to 0

Parallelizable Key Recovery Framework

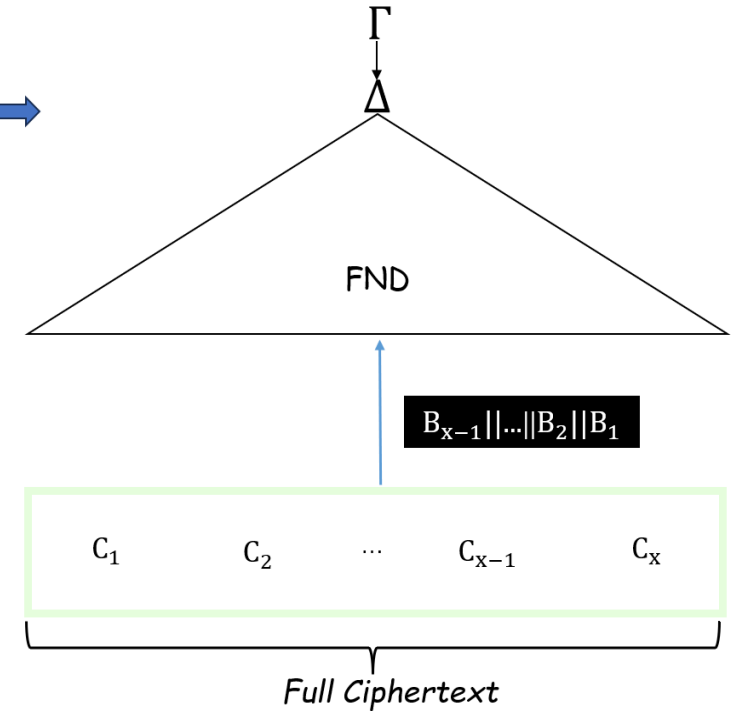
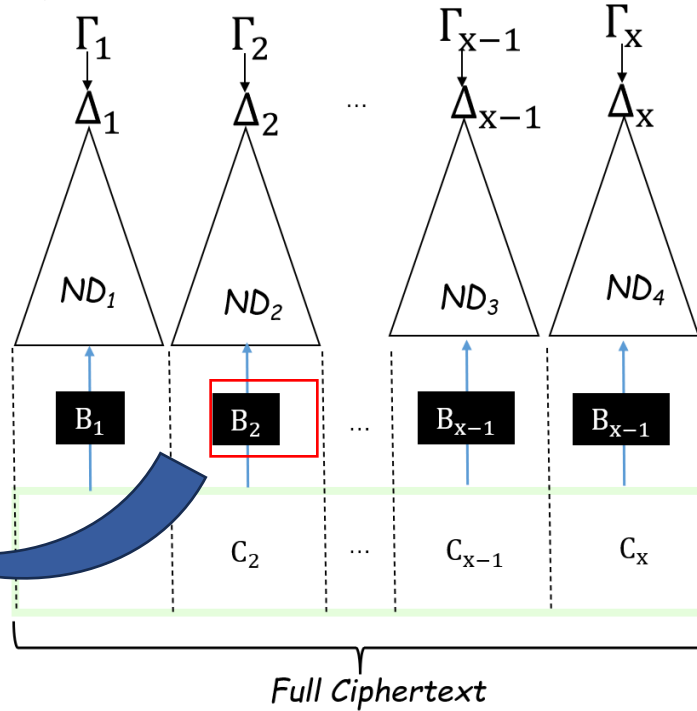
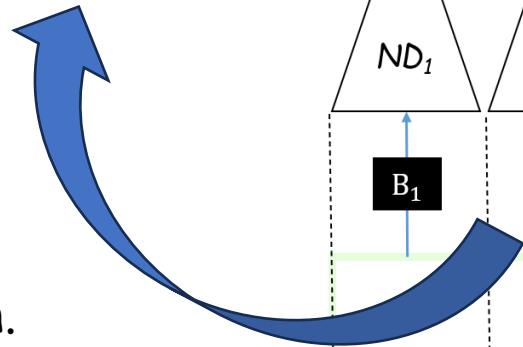


➤ Advantages of the new framework:

Do not need to be concatenated with the n keys recommended in the previous step.



The time is reduced to $1/n$.



Under the condition of multiple GPUs, all distinguisher perform key recovery at the same time



The time is reduced to $1/x$

➤ Application to SPECK64/96 and SPECK96/96 :

Target	Round	Avg. Time ¹	hw(kg, rk) ²	Configure	Reference
SPECK64/96	9	66s	1.93	$1 + 1r_{CD} + 6r_{ND}$	Chen et al.
		29s	0.80	$1 + 1r_{CD} + 6r_{ND}$	This paper
SPECK96/96	10	303s	1	$1 + 1r_{CD} + 7r_{ND}$	Chen et al.
		163s	0.44	$1 + 1r_{CD} + 7r_{ND}$	This paper

¹ Avg. Time refers to the average time required to perform a single attack on a computer equipped with an NVIDIA 3060 graphics card.

² hw(kg, rk) refers to the average Hamming distance between the guessed key and the actual key

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- The longest-round traditional analysis results for SPECK64/96 and SPECK96/96 [Song et al. in 2016]

Target	Rounds attacked/Total rounds	Reference
SPECK64/96	19/26	Song et al.
SPECK96/96	20/28	

- The Differential-Neural Cryptanalysis results for SPECK64/96 and SPECK96/96

Target	Round	Configure	Reference
SPECK64/96	9	$1 + 1r_{CD} + 6r_{ND}$	Chen et al.
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SPECK96/96	10	$1 + 1r_{CD} + 7r_{ND}$	Chen et al.
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➤ Selection of Input Differences

Gohr choose $0x0040/0000$ as the input difference



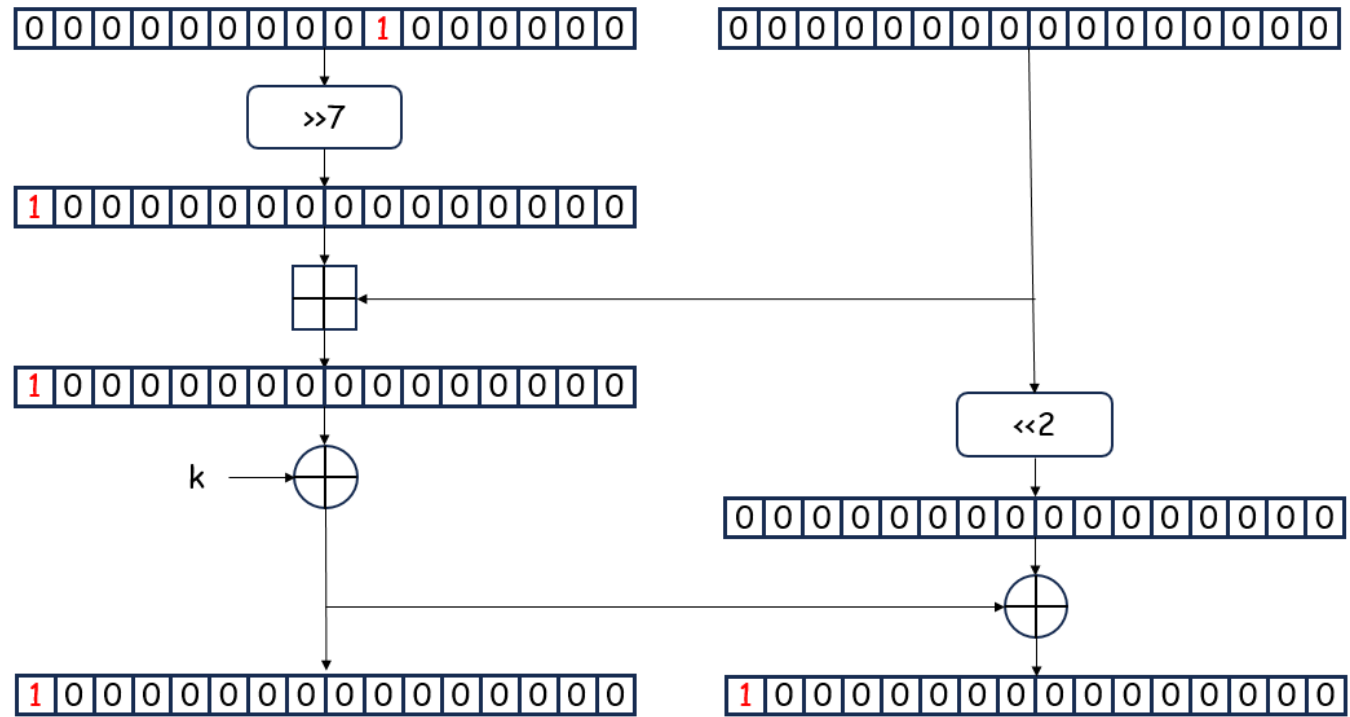
The carry effect produced by the first round of modular addition is eliminated.



$0x0040/0000 \xrightarrow{1} 0x8000/8000$



Making the difference distribution more concentrated is beneficial to the neural network in learning ciphertext knowledge.



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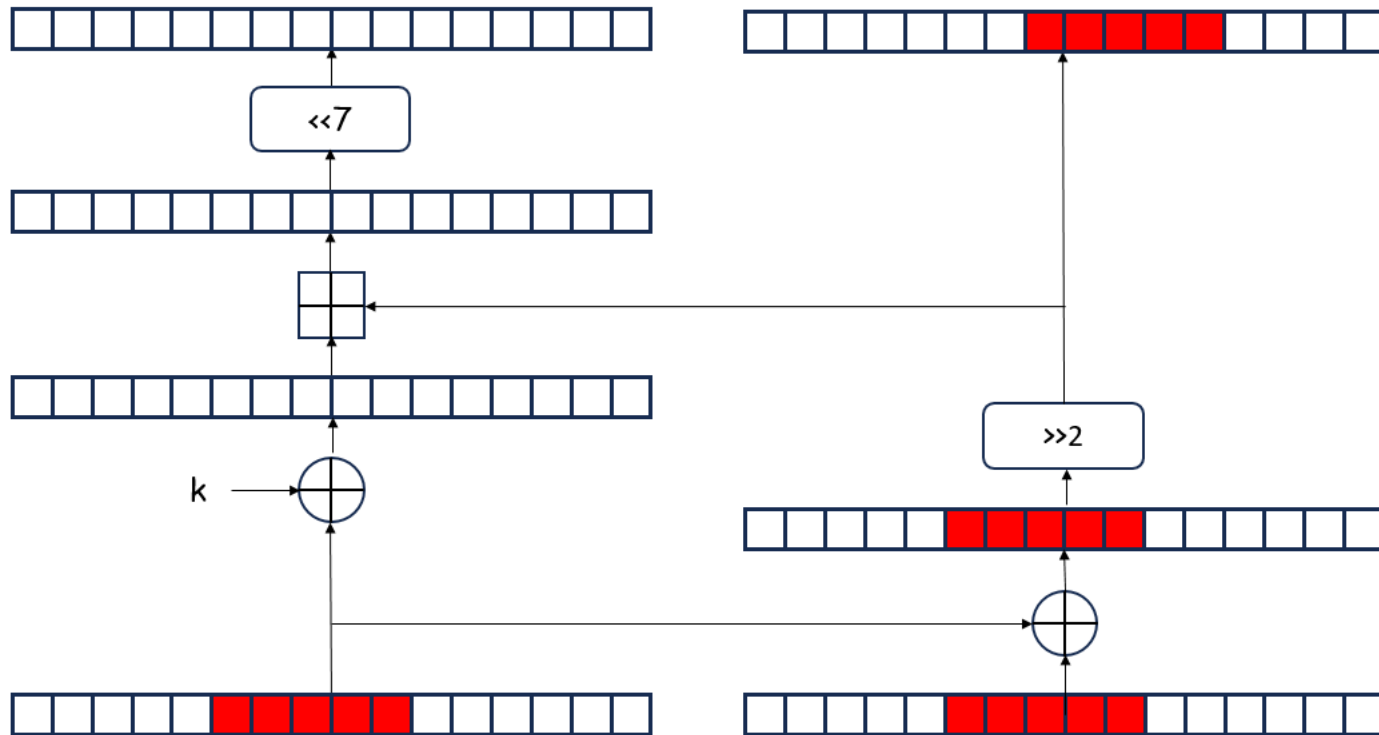


Making the difference distribution more concentrated is
beneficial to the neural network in learning ciphertext knowledge.

We choose $(0x80, 0)$, $(0x8000, 0)$, and
 $(0x800000, 0)$ as the input differences
for the neural network of SEPCK64/96.



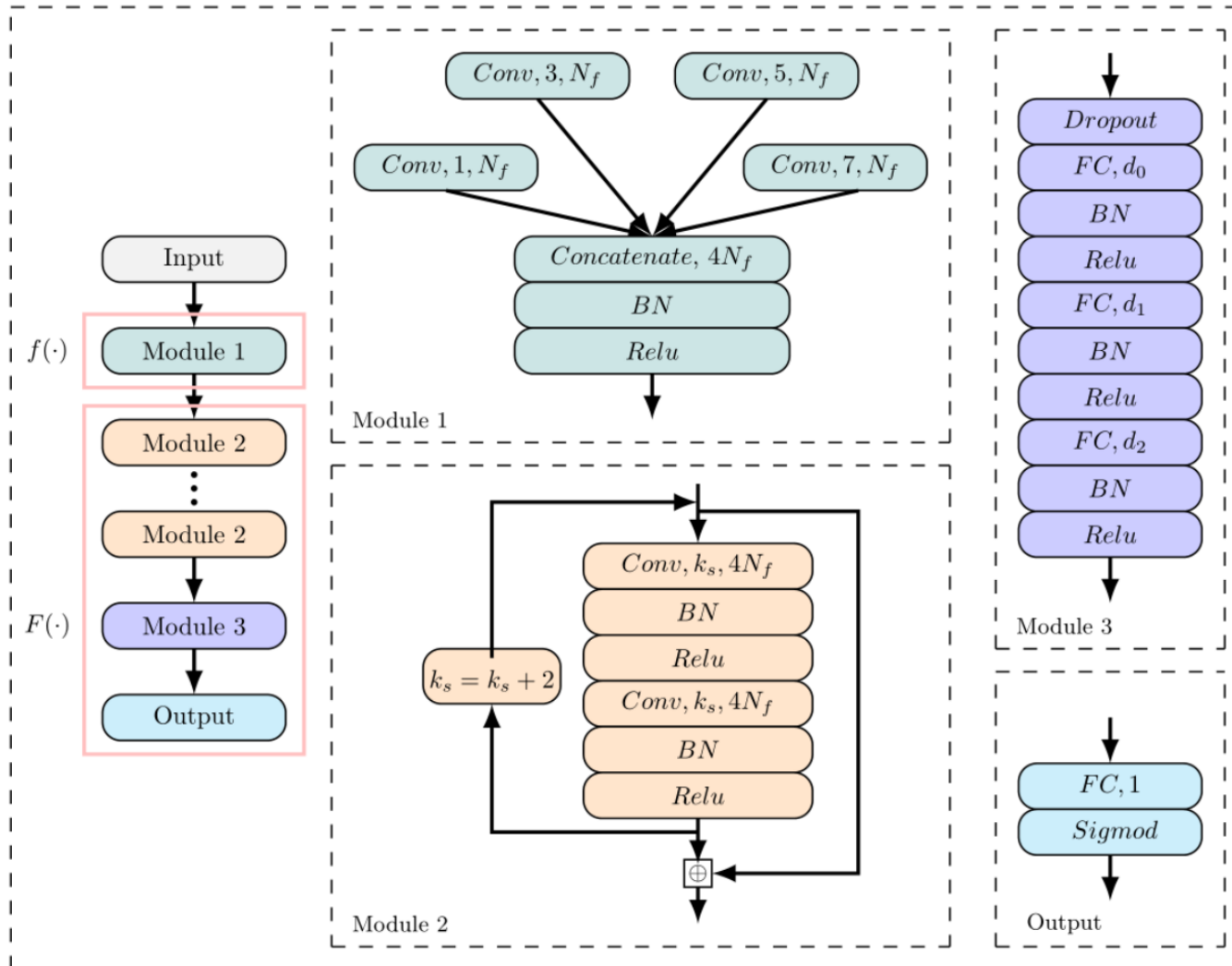
- Including information from the penultimate round and multiple pairs.



With ciphertext fragments, we can calculate the ciphertext fragments of the right half of the penultimate round without knowing the key. Add them to training data.

At the same time, we include ciphertext information from multiple different ciphertext pairs in one training sample.

➤ Network Architecture



Using the network structure proposed by Zhang et al. and use a staged training method.

➤ Neural distinguishers against 7-round SPECK64

Distinguisher	Input difference	Ciphertext bits	Accuracy
ND ₁	(0x8000, 0),	{17 ~ 8}	0.613
ND ₂	(0x80, 0)	{29 ~ 18}	0.662
ND ₃	(0x800000, 0),	{31, 30, 7 ~ 0}	0.620

➤ Result of Key Recovery Attack for 10-Round SPECK64/96

Target	Round	Avg. Time ¹	hw(kg, rk) ²	Configure	Reference
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	10	424s	1.89	$1 + 1r_{CD} + 7r_{ND}$	This paper
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- A parallelizable multi-stage key recovery framework based on neural distinguishers, accelerating the key recovery attack on largestate block ciphers.
- Enhance the ability of the distinguisher in the multi-stage recovery framework.



Thank you!

Q & A