# EasyBC: A Cryptography-Specific Language for Security Analysis of Block Ciphers against Differential Cryptanalysis 

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#### Abstract

Differential cryptanalysis is a powerful algorithmic-level attack, playing a central role in evaluating the security of symmetric cryptographic primitives. In general, the resistance against differential cryptanalysis can be characterized by the maximum expected differential characteristic probability. In this paper, we present generic and extensible approaches based on mixed integer linear programming (MILP) to bound such probability. We design a high-level cryptography-specific language EASYBC tailored for block ciphers and provide various rigorous procedures as differential denotational semantics, to automate the generation of MILP from block ciphers written in EasyBC. We implement an open-sourced tool that provides support for fully automated resistance evaluation of block ciphers against differential cryptanalysis. The tool is extensively evaluated on 23 real-life cryptographic primitives including all the 10 finalists of the NIST lightweight cryptography standardization process. The experiments confirm the expressivity of EAsyBC and show that the tool can effectively prove the resistance against differential cryptanalysis for all block ciphers under consideration. EAsyBC makes resistance evaluation against differential cryptanalysis easily accessible to cryptographers.


## CCS Concepts: • Theory of computation; • Security and privacy;

Additional Key Words and Phrases: Cryptography-Specific Language, Block Ciphers, Differential Cryptanalysis

## ACM Reference Format:

Pu Sun, Fu Song, Yuqi Chen, and Taolue Chen. 2024. EasyBC: A Cryptography-Specific Language for Security Analysis of Block Ciphers against Differential Cryptanalysis. Proc. ACM Program. Lang. 8, POPL, Article 29 (January 2024), 34 pages. https://doi.org/10.1145/3632871

## 1 INTRODUCTION

A block cipher is a symmetric cryptographic technique that uses the same key to encrypt and decrypt data in fixed-size blocks. Guided by the design principles of block ciphers [Shannon 1949], a vast number of block ciphers have been proposed, varying in, e.g., network structures and block sizes. Many of them have been standardized and are widely used in daily life to provide confidentiality, integrity, and authentication. Differential cryptanalysis, proposed by [Biham and Shamir 1990], is a powerful algorithmic-level attack against block ciphers by analyzing the effect of particular differences in input pairs on the differences of pairs of intermediate states under

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ACM 2475-1421/2024/1-ART29
https://doi.org/10.1145/3632871
the same key. It has proved to be a very effective attack, which has broken block ciphers such as DES [Biham and Shamir 1990], FEAL [Aoki et al. 1997], WARP [Teh and Biryukov 2022], and K-Cipher [Mahzoun et al. 2022]. As a result, a provable guarantee of the resistance of block ciphers against differential cryptanalysis has become a standard criterion for new block ciphers and indeed a basic requirement for them to be standardized [Katz and Lindell 2014]. In light of the diversity and wide deployment of block ciphers, a generic, and ideally automated, approach which can be used to evaluate their resistance against differential cryptanalysis, becomes indispensable.

In general, the resistance of block ciphers against differential cryptanalysis is commonly characterized by the maximum expected differential characteristic probability (MaxEDCP for short, the definition of which is fairly standard but technical, and will be given in Section 2.1). A block cipher is considered to be resistant against differential cryptanalysis if the MaxEDCP is no greater than $2^{-\epsilon}$, where $\mathscr{b}$ is the block size of the block cipher [Heys 2002; Lai et al. 1991]. In light of this, the central task of security analysis for block ciphers against differential cryptanalysis is reduced to computing such probability. To this end, [Matsui 1994] proposed a branch-and-bound searching algorithm that traverses differential characteristics in a depth-first manner and computes their probabilities during traversal. However, it becomes inefficient with the increasing of candidate differential characteristics. Various heuristics are then proposed to improve the efficiency [Aoki et al. 1997; Bao et al. 2014; Biryukov and Nikolić 2010; Ji et al. 2021], but they harness cipher-specific optimizations and thus require sophisticated programming skills.

Several alternative methods are introduced, which reduce to mixed integer linear program (MILP [Mouha et al. 2011]), Boolean satisfiability problem (SAT [Mouha and Preneel 2013]), or satisfiability modulo theory (SMT [Aumasson et al. 2014]). These methods allow cryptanalysts to specify the problem for each cipher using the input language of MILP/SAT/SMT solvers so that respective solvers can be harnessed. Plenty of modeling methods for cryptography-specific operations such as substitution-box (S-box), Exclusive-OR (XOR), and linear transformations, as well as heuristics, are proposed to improve efficiency and accuracy. However, insofar cryptanalysts have to manually model each cipher in the tool they choose to use, or at best write a model generation script for each cipher, which is usually intricate, error-prone, and laborious, as it requires the cryptanalysts to be familiar with the specific tool and a wide range of modeling methods. There appears to be a lack of language support, unified computational approaches, and fully automated tools for evaluating the resistance of block ciphers against differential cryptanalysis.

Contributions. In this work, our primary aim is to develop a generic and automated approach for evaluating the resistance of block ciphers against differential cryptanalysis. To achieve this goal, we begin by designing a novel high-level statically-typed, C-like cryptography-specific language, EAsyBC (Easy Block Cipher), tailored for block ciphers. Besides standard types and operations, EAsyBC provides cryptography-specific types (e.g., S-boxes, P-boxes) and operations (e.g., substitution via S-box, linear transformation via P-box, or matrix-vector product), to facilitate the implementation of block ciphers. (Note that an S-box takes $m$ input bits and transforms them into $n$ output bits and P-box, short for Permutation-box, is an array specifying a permutation of inputs.) The language is fully specified by a formal grammar together with typing rules and operational semantics, enabling further automatic analysis.

Concretely speaking, the analysis of block cipher resistance against differential cryptanalysis primarily involves computing the MaxEDCP. However, calculating MaxEDCP precisely is practically infeasible. As a result, we employ a strategy to calculate a tight upper bound that is sufficient to demonstrate resistance. Specifically, we adopt the typical MILP-based approach where linear constraints are used to characterize the dependency (i.e., feasibility) between input and output differences of each operation, and the optimization objective is to minimize an upper bound. In
particular, we give a rigorous procedure, formalized as a differential denotational semantics, to automate the generation of MILP from EAsyBC programs, which not only unifies and optimizes the existing but also discovers new generation processes. Our approach is of generic nature, thanks to the expressivity of the EAsyBC langauge, namely, a multitude of block ciphers can be handled in a unified way.

Technically, the generation of MILP can be done either at the word or bit level. The former is less involved and generates fewer constraints, but is limited to certain block ciphers; the latter approach is more fine-grained and has wider applicability. In both cases, the general strategy is to determine the lower bound of the minimum number of (differentially) active S-boxes, i.e., S-boxes whose input differences are nonzero under two executions [Biryukov and Nikolić 2010; Heys 2002], from which the upper bound of the MaxEDCP can be deduced according to [Heys 2002; Sun et al. 2014a]. While this strategy is efficient, it is important to note that the obtained upper bound may not always be sufficiently tight and may not be applicable for certain ciphers. To address this limitation, we introduce an extended bit-wise approach that directly bounds the MaxEDCP by encoding probabilities using additional Boolean variables. We have successfully implemented our approach as the first fully automated tool for evaluating the resistance of block ciphers against differential cryptanalysis. This tool eliminates the need for cryptanalysts to possess knowledge of MILP generation for cryptographic operations. Instead, they can simply write a program in EAsyBC for a block cipher. Moreover, the generation of MILP from EAsyBC program is modular, i.e., each cryptographic operation in EASYBC is associated with its own MILP generation rule and new generation rules could be easily added by implementing designated APIs. As a result, our approach exhibits excellent extensibility for new block ciphers, enabling a wider range of applicability.

To evaluate the tool, we implement 23 realistic cryptographic primitives with EAsYBC, including all the 10 finalists of the NIST lightweight cryptography standardization process [NIST 2023] and other commonly used block ciphers, covering both substitution-permutation network (SPN) based ciphers (e.g., AES, PRESENT, and GIFT) and balanced Feistel networks (BFN) based ciphers (e.g., DES, LBLOCK, and TWINE). It turns out that EAsYBC can express these block ciphers in a considerably more succinct way, demonstrating our language's expressiveness. Moreover, it turns out that our tool is able to effectively handle all realistic block ciphers under consideration, showcasing its capability in proving the resistance of block ciphers against differential cryptanalysis.

In addition, we compare various alternative MILP generation methods for cryptographic operations. Interestingly, we observe that certain methods may generate fewer constraints and variables. However, it is worth noting that such reductions may actually have a negative impact on the overall MILP-solving process. For instance, the recent S-box modeling method proposed by [Udovenko 2021] produces the fewest constraints, but also exhibits the least performance to solve those constraints. (Overall, it is significantly less efficient than some alternatives.) Our findings shed light on the selection of generation methods among various alternatives for practical applications.

We summarize the main contributions as follows.

- We design a high-level cryptography-specific language EAsyBC tailored for block ciphers, enabling further automatic analysis.
- We give generic and extensible approaches for automated resistance evaluation of block ciphers written in EasyBC against differential cryptanalysis.
- We implement and extensively evaluate an open-sourced prototype of EASYBC, which confirms the expressiveness of EAsyBC and the effectiveness of our approach.

Structure. The rest of the paper is organized as follows. Section 2 presents the background of block ciphers and differential cryptanalysis. Section 3 introduces EAsyBC and an overview of our approach. Section 4 presents three key utilities used in our MILP generation. Section 5 and Section 6

Proc. ACM Program. Lang., Vol. 8, No. POPL, Article 29. Publication date: January 2024.
describe the word-wise and bit-wise approach. Section 7 describes the extended bit-wise approach. Section 8 reports the experimental results. We discuss related work in Section 9 and conclude this work in Section 10.

## 2 BACKGROUND

Throughout this paper, $\mathbb{B}$ denotes the Boolean domain $\{0,1\}$, and $\mathbb{N}$ denotes the set of non-negative integers. Boolean values are treated as integers in arithmetic computations. Given a vector/array $\vec{x}$, $\vec{x}_{i}$ denotes the $(i+1)$-th entry. Given a matrix $M, M_{i}$ denotes the $(i+1)$-th row, and $M_{i, j}$ denotes the $(j+1)$-th entry of $M_{i}$. An $m$-bitstream is Boolean vector $\vec{b}$ with $m$ entries. We denote by $\oplus$ the bit-wise XOR operator. $\vec{b} \| \vec{b}^{\prime}=\left(b_{0}, \cdots, b_{m}, b_{0}^{\prime}, \cdots, b_{n}^{\prime}\right)$ is the concatenation of two bitstreams $\vec{b}=\left(b_{0}, \cdots, b_{m}\right)$ and $\vec{b}^{\prime}=\left(b_{0}^{\prime}, \cdots, b_{n}^{\prime}\right)$. We denote by $\operatorname{bin}(x)$ the binary representation of an unsigned integer $x$ as a bitstream.

### 2.1 Block ciphers

Block ciphers are a type of symmetric cryptography, which encrypts and decrypts data in fixed-size (e.g., 64 or 128 bits) blocks using the same key [Bogdanov 2010; Knudsen 1997].

Definition 2.1. A block cipher is a function Enc: $\mathbb{B}^{A} \times \mathbb{B}^{a} \rightarrow \mathbb{B}^{a}$ such that for every key $K \in \mathbb{B}^{/}$, $\operatorname{Enc}(K, \cdot)$ is a bijective function, where $b$ is the block size and $\nless$ is the key size.

Intuitively, $\operatorname{Enc}(K, \cdot)$ is a keyed-permutation that maps an input block to an output block, where a block is a $\ell$-bitstream, the key $K$ determines which permutation to perform. The input and output of $\operatorname{Enc}(K, \cdot)$ are called plaintext and ciphertext, respectively. A block cipher Enc is ideal if it is defined by assigning a uniformly drawn permutation to each of the $2^{\hbar}$ keyed-permutations. An ideal block cipher is commonly considered to be computationally secure if the key size $\ell$ is large enough since the brute-force attack requires $O\left(2^{/ /}\right)$time. However, it is extremely difficult to implement an ideal block cipher for practical block sizes (e.g., 64 or 128), as one randomly drawn permutation $\operatorname{Enc}(K, \cdot)$ has to be stored for each given key $K$.

To be efficient yet strong, modern block ciphers apply several (possibly distinct) keyed permutations, where one keyed permutation is a round and implemented by a round function.

Definition 2.2. A $\mu$-round iterative block cipher $(\mu-\mathrm{IBC})$ is a function Enc: $\mathbb{B}^{\natural} \times \mathbb{B}^{a} \rightarrow \mathbb{B}^{a}$ such that for every key $K \in \mathbb{B}^{n}$,

$$
\operatorname{Enc}(K, \cdot)=\operatorname{Enc}_{\mu}\left(K^{\mu}, \cdot\right) \circ \cdots \circ \operatorname{Enc}_{1}\left(K^{1}, \cdot\right),
$$

where $\mathrm{Enc}_{i}: \mathbb{B}^{h} \times \mathbb{B}^{6} \rightarrow \mathbb{B}^{6}$ is the $i$-th round with its subkey $K^{i}$ for $1 \leq i \leq r$, symbol $\circ$ denotes function composition, and the subkeys are generated via a key schedule algorithm $g: \mathbb{B}^{k} \rightarrow\left(\mathbb{B}^{/ /}\right)^{\kappa}$, i.e., $g(K)=\left(K^{1}, \cdots, K^{\mu}\right)$.

Given a key $K \in \mathbb{B}^{\ell}, \operatorname{Enc}(K, \cdot)$ is used for encryption and $\operatorname{Dec}(K, \cdot)$ is used for decryption, i.e., $\operatorname{Dec}(K, \cdot)=\operatorname{Enc}_{1}^{-1}\left(K^{1}, \cdot\right) \circ \cdots \circ \operatorname{Enc}_{\mu}^{-1}\left(K^{\kappa}, \cdot\right)$, where $K^{i}$ is the $i$-th round subkey in Definition 2.2.

Block ciphers can be built in various ways following iterative cipher schemes. The two most widely used are substitution-permutation networks (SPN) [Kam and Davida 1979] and balanced Feistel networks (BFN) [Nyberg 1996]. Note that BFN is used in the former U.S. encryption standard (DES-Data Encryption Standard) [Fox 2000] and SPN is used in the current one (AES-Advanced Encryption Standard) [Daemen and Rijmen 1999]. A brief introduction of BFN and SPN is given in [Sun et al. 2023, Section A].
In the sequel, we fix a $\kappa$-IBC Enc : $\mathbb{B}^{\ell} \times \mathbb{B}^{6} \rightarrow \mathbb{B}^{6}$ with rounds $\mathrm{Enc}_{1}, \cdots, \mathrm{Enc}_{\mu}$. We assume that the attacker knows all dethe tails of the encryption and decryption except for the secret key.

### 2.2 Differential Cryptanalysis

Differential cryptanalysis recovers the secret key by exploiting the fact that the probability of some output differences of rounds in a non-ideal block cipher is higher than the expected value (i.e., $2^{-\mathscr{d}}$ ) for certain input differences [Biham and Shamir 1990]. We review the related concepts below.
Difference and differential. Given two $b$-bitstreams $X \in \mathbb{B}^{d}$ and $X^{\prime} \in \mathbb{B}^{b}$, their (XOR-) difference $\Delta X$ is defined by $\Delta X=X \oplus X^{\prime}$. Note that for any fixed difference $\Delta X$, there are exactly $2^{\star}$ pairs ( $X, X^{\prime}$ ) such that $X \oplus X^{\prime}=\Delta X$. A differential is defined to be a pair of differences $(\Delta X, \Delta Y)$.

Given a pair of inputs $\left(X, X^{\prime}\right) \in \mathbb{B}^{n} \times \mathbb{B}^{n}$ for a (deterministic) function $f: \mathbb{B}^{n} \rightarrow \mathbb{B}^{m}$, the input difference of the function $f$ is the difference $\Delta X=X \oplus X^{\prime}$ of the inputs ( $X, X^{\prime}$ ), and the output difference of $f$ is the difference $f(X) \oplus f\left(X^{\prime}\right)$ of the outputs $\left(f(X), f\left(X^{\prime}\right)\right)$. Clearly, when the input difference is fixed to be $\Delta X$, the output difference of an input $X$ is $f(X) \oplus f(X \oplus \Delta X)$.

The probability $\operatorname{Pr}_{f}(\Delta X, \Delta Y)$ of a given differential $(\Delta X, \Delta Y)$ for the function $f$ is the proportion of inputs $X \in \mathbb{B}^{n}$ such that the output difference $f(X) \oplus f(X \oplus \Delta X)$ is equal to $\Delta Y$, i.e.,

$$
\operatorname{Pr}_{f}(\Delta X, \Delta Y)=\frac{\left|\left\{X \in \mathbb{B}^{n} \mid f(X) \oplus f(X \oplus \Delta X)=\Delta Y\right\}\right|}{2^{n}}
$$

In particular, for an $i$-th round $\mathrm{Enc}_{i}: \mathbb{B}^{\hbar} \times \mathbb{B}^{d} \rightarrow \mathbb{B}^{\natural}$, with fixed subkey $K^{i}, \operatorname{Pr}_{\operatorname{Enc}_{i}\left(K^{i},\right)}\left(\Delta X^{i-1}, \Delta X^{i}\right)$ is the probability of a differential $\left(\Delta X^{i-1}, \Delta X^{i}\right)$ for the function $\operatorname{Enc}_{i}\left(K^{i}, \cdot\right)$.

It is worth mentioning that in differential cryptanalysis, a key assumption is that the output difference $\mathrm{Enc}_{i}\left(K^{i}, X^{i-1}\right) \oplus \mathrm{Enc}_{i}\left(K^{i}, X^{i-1} \oplus \Delta X^{i-1}\right)$ of the $i$-th round is independent of the subkey $K^{i}$ for any fixed input $X^{i-1}$ and input difference $\Delta X^{i-1}$. As a result, for clarity, $\operatorname{Pr}_{\operatorname{Enc}_{i}\left(K^{i},\right)}(\cdot)$ is simply written as $\operatorname{Pr}_{\mathrm{Enc}_{i}}(\cdot)$.
Differential characteristic. An $s$-round differential characteristic is a vector $\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ of differences, where

- $\Delta X^{0}$ a nonzero input difference to the cipher Enc : $\mathbb{B}^{k} \times \mathbb{B}^{d} \rightarrow \mathbb{B}^{a}$,
- for $1 \leq i \leq s,\left(\Delta X^{i-1}, \Delta X^{i}\right)$ is a differential of the $i$-th round $\operatorname{Enc}_{i}\left(K^{i}, \cdot\right)$.

Definition 2.3. The differential characteristic probability $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ of an $s$-round differential characteristic $\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ is the proportion of inputs $X^{0} \in \mathbb{B}^{t}$ to the cipher Enc such that the output difference of the $i$-th round $\mathrm{Enc}_{i}$ is $\Delta X^{i}$ for every $1 \leq i \leq s$ in the two executions of Enc under the two inputs ( $K, X^{0}$ ) and ( $K, X^{0} \oplus \Delta X^{0}$ ), namely,

$$
\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)=\frac{\left|\left\{X^{0} \in \mathbb{B}^{\epsilon} \mid \forall i .1 \leq i \leq s . \operatorname{Enc}_{\leq i}\left(X_{0}\right) \oplus \mathrm{Enc}_{\leq i}\left(X_{0} \oplus \Delta X^{0}\right)=\Delta X^{i}\right\}\right|}{2^{6}}
$$

where $\operatorname{Enc}_{\leq i}:=\operatorname{Enc}_{i}\left(K^{i}, \cdot\right) \circ \cdots \circ \operatorname{Enc}_{1}\left(K^{1}, \cdot\right)$ for $1 \leq i \leq s$.
Recall that we assumed that the output difference $\mathrm{Enc}_{i}\left(K^{i}, X^{i-1}\right) \oplus \mathrm{Enc}_{i}\left(K^{i}, X^{i-1} \oplus \Delta X^{i-1}\right)$ of the $i$-th round is independent upon the subkey $K^{i}$ for any fixed input $X^{i-1}$ and input difference $\Delta X^{i-1}$. Thus, the probability $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ does not depend on the $s$-round subkey $K^{s}$, but depends on the other subkeys $K^{1}, \cdots, K^{s-1}$. It is known that $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ is upper bound by $\prod_{i=1}^{s} \operatorname{Pr}_{\mathrm{Enc}_{i}}\left(\Delta X^{i-1}, \Delta X^{i}\right)$ [Heys and Tavares 1996], and they are the same if Enc is a Markov cipher and its round subkeys are independent [Lai et al. 1991].
In a resistant block cipher, for any fixed key $K$, the probability $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ should be small enough for any differential characteristic ( $\Delta X^{0}, \cdots, \Delta X^{s}$ ) if the input $X^{0}$ is sampled uniformly. However, in practice, $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}\left(\Delta X^{0}, \cdots, \Delta X^{s}\right)$ may be higher than $2^{-t}$ based on which an attacker can efficiently recover the subkey $K^{s+1}$. The differential characteristic ( $\widetilde{\Delta X}^{0}, \cdots, \widetilde{\Delta X}^{s}$ ) is said to be optimal if it attains the greatest expected differential characteristic probability among all the $s$ round differential characteristics. We remark that optimal differential characteristics are commonly assumed to be identical for different keys $K$ during the attack, as the actual key is unknown to
the adversary before attacking. In the sequel, for simplicity, $\operatorname{Pr}_{\operatorname{Enc}(K, \cdot)}(\cdot)$ is written as $\operatorname{Pr}_{\text {Enc }}(\cdot)$ and $\operatorname{Pr}_{\text {Enc }}\left(\widetilde{\Delta X}^{0}, \cdots, \widetilde{\Delta X}^{s}\right)$ is refer to as the maximum expected differential characteristic probability (MaxEDCP). The key recovering procedure is given in [Sun et al. 2023, Section B], where the number of plaintexts required to infer the ( $s+1$ )-round subkey $K^{s+1}$ is proportional to $\frac{1}{\operatorname{Pr}_{\text {Enc }}\left(\widetilde{\Delta X}^{0}, \ldots, \widetilde{\Lambda X}^{s}\right)}$ [Heys 2002], i.e., the reciprocal of the MaxEDCP of $s$-round differential characteristics. As a result, if one could show that an upper bound of $\operatorname{Pr}_{\text {Enc }}\left(\widetilde{\Delta X}^{0}, \cdots, \widetilde{\Delta X}^{s}\right)$ is no greater than $2^{-6}$, one would conclude the resistance of the block cipher against such differential cryptanalysis.

### 2.3 Active S-box, Differential Distribution Table and Branch Number

We introduce some notions of active S-box, differential distribution table and branch number which will be used in our approach.
Active S-boxes. The number of active S-boxes can be used to upper bound the MaxEDCP. Assume S-boxes are distinct in the cipher Enc.

Definition 2.4. [Heys 2002] Given a key $K$, an input $X^{0}$ and an input difference $\Delta X^{0}$ to the cipher Enc, an S-box $\mathcal{S}$ is active if the two inputs to $\mathcal{S}$ are distinct in the two executions of the cipher Enc under two inputs ( $K, X^{0}$ ) and ( $K, X^{0} \oplus \Delta X^{0}$ ), otherwise it is inactive.

We denote by $\mathcal{N}_{\text {diff }}$ the minimum number of the active $S$-boxes in all the possible pairs of executions. The probability of optimal $s$-round differential characteristics is bounded from above by $p^{N_{\text {diff }}}$ [Heys 2002], where $p$ denotes the maximum probability $\operatorname{Pr} \operatorname{Pr}_{\mathcal{S}}(\Delta X, \Delta Y)$ among all the nonzero differentials $(\Delta X, \Delta Y)$ for any $S$-box $\mathcal{S}$ which is active in the $s$-round differential characteristics.
Differential distribution table. A differential distribution table (DDT) is a data structure to represent the distribution $\operatorname{Pr}_{f}$ of a function $f$ for all possible differentials. It also explicitly expresses the dependency (i.e., feasibility) between input and output differences of the function $f$.

Definition 2.5. Given a function $f: \mathbb{B}^{n_{1}} \times \cdots \times \mathbb{B}^{n_{i}} \rightarrow \mathbb{B}^{m_{1}} \times \cdots \times \mathbb{B}^{m_{j}}$, its DDT $\mathcal{D}_{f}$ is table such that for every vector of input differences $\left(\Delta X^{1}, \cdots, \Delta X^{i}\right)$ and every vector of output differences $\left(\Delta Y^{1}, \cdots, \Delta Y^{j}\right)$, the entry $\mathcal{D}_{f}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ gives the number of vectors of inputs $\left(X^{1}, \cdots, X^{i}\right) \in \mathbb{B}^{n_{1}} \times \cdots \times \mathbb{B}^{n_{i}}$ such that

$$
f\left(X^{1}, \cdots, X^{i}\right) \oplus f\left(X^{1} \oplus \Delta X^{1}, \cdots, X^{i} \oplus \Delta X^{i}\right)=\left(\Delta Y^{1}, \cdots, \Delta Y^{j}\right) .
$$

The probability $\operatorname{Pr}_{f}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ can be deduced from the DDT $\mathcal{D}_{f}$ :

$$
\operatorname{Pr}_{f}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)=\frac{\mathcal{D}_{f}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)}{2^{n_{1}+\cdots+n_{i}}}
$$

We say the the vector of input and output differences ( $\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}$ ) is feasible for $f$, if the probability $\operatorname{Pr}_{f}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ is nonzero, otherwise it is infeasible. When the input space of the function $f$ is small, its DDT $\mathcal{D}_{f}$ can be computed by enumeration.

Example 2.6. Consider the AND operation $\wedge: \mathbb{B} \times \mathbb{B} \rightarrow \mathbb{B}$. We have that

$$
\Delta Y=\left(X^{1} \wedge X^{2}\right) \oplus\left(\left(X^{1} \oplus \Delta X^{1}\right) \wedge\left(X^{2} \oplus \Delta X^{2}\right)\right)
$$

The DDT $\mathcal{D}_{\wedge}$ is shown in Table 1, e.g., $\operatorname{Pr}_{\wedge}(0,0,0)=\frac{4}{4}=1$, and $\operatorname{Pr}_{\wedge}\left(\Delta X^{1}, \Delta X^{2}, \Delta Y\right)=\frac{2}{4}=\frac{1}{2}$ if $\Delta X^{1}=1$ or/and $\Delta X^{2}=1$ for any fixed $\Delta Y$.

The DDT $\mathcal{D}_{\vee}$ of the OR operation $\vee: \mathbb{B} \times \mathbb{B} \rightarrow \mathbb{B}$ is the same as $\mathcal{D}_{\wedge}$. We can observe that the input difference $(0,0)$ cannot lead to the output difference 0 for both the AND and OR operations.

Branch number. The (differential) branch number of a function is also used to characterize the dependency between input and output differences of the function [Daemen and Rijmen 1999].

Table 1. The $\operatorname{DDT}\left(\mathcal{D}_{\wedge}, \mathcal{D}_{\vee}\right)$ for $\wedge / \vee: \mathbb{B} \times \mathbb{B} \rightarrow \mathbb{B}$.

| $\left(\Delta X^{1}, \Delta X^{2}\right)$ | $(0,0)$ | $(0,1)$ | $(1,0)$ | $(1,1)$ |
| :--- | :---: | :---: | :---: | :---: |
| $\Delta Y=0$ | 4 | 2 | 2 | 2 |
| $\Delta Y=1$ | 0 | 2 | 2 | 2 |

Table 2. Branch numbers of $+,-, \wedge, \vee, \oplus$.

|  | + | - | $\wedge$ | $\vee$ | $\oplus$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathcal{B}_{\mathrm{ww}}^{\min }$ and $\mathcal{B}_{\mathrm{bw}}^{\min }$ | 2 | 2 | 1 | 1 | 2 |
| $\mathcal{B}_{\mathrm{ww}}^{\max }$ | 3 | 3 | 3 | 3 | 3 |
| $\mathcal{B}_{\mathrm{bw}}^{\max }$ | $3 \mathrm{n}-1$ | $3 \mathrm{n}-1$ | 3 n | 3 n | 2 n |

Definition 2.7. Given a function $f: \mathbb{B}^{n_{1}} \times \cdots \times \mathbb{B}^{n_{i}} \rightarrow \mathbb{B}^{m_{1}} \times \cdots \times \mathbb{B}^{m_{j}}$, its minimum (resp. maximum) word-wise branch number $\mathcal{B}_{\mathrm{ww}}^{\min }(f)$ (resp. $\left.\mathcal{B}_{\mathrm{ww}}^{\max }(f)\right)$ is defined as

```
\(\mathcal{B}_{\mathrm{ww}}^{\max }(f)=\max \left\{\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right) \mid \forall 1 \leq \ell \leq n . X^{\ell}, \Delta X^{\ell} \in \mathbb{B}^{n_{\ell}} . \operatorname{BNCond}(f)\right\}\)
\(\mathcal{B}_{\mathrm{ww}}^{\min }(f)=\min \left\{\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right) \mid \forall 1 \leq \ell \leq n . X^{\ell}, \Delta X^{\ell} \in \mathbb{B}^{n_{\ell}}\right.\). BNCond \(\left.(f)\right\}\)
```

where $\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ counts the number of nonzero entries in the vector $\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ and the branch-number condition $\operatorname{BNCond}(f)$ is:

$$
\operatorname{BNCond}(f)=\binom{\left(\Delta X^{1} \neq 0 \vee \cdots \vee \Delta X^{i} \neq 0\right) \wedge\left(Y^{1}, \cdots, Y^{j}\right)=f\left(X^{1}, \cdots, X^{i}\right)}{\wedge\left(Y^{1} \oplus \Delta Y^{1}, \cdots, Y^{j} \oplus \Delta Y^{j}\right)=f\left(X^{1} \oplus \Delta X^{1}, \cdots, X^{i} \oplus \Delta X^{i}\right)}
$$

Likewise, the minimum (resp. maximum) bit-wise branch number $\mathcal{B}_{\mathrm{bw}}^{\min }(f)$ (resp. $\mathcal{B}_{\mathrm{bw}}^{\max }(f)$ ) of the function $f$ is defined except that $\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ counts the number of 1 bits in the bitstream $\Delta X^{1}\|\cdots\| \Delta X^{i}\left\|\Delta Y^{1}\right\| \cdots \| \Delta Y^{j}$, i.e., Hamming weight.

Intuitively, when the input differences of the function $f$ are not all 0 bits (i.e., $\Delta X^{1} \neq 0 \vee \cdots \vee \Delta X^{i} \neq$ 0 ), the number of nonzero entries in any input and output differences $\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ of the function $f$ ranges from $\mathcal{B}_{\mathrm{ww}}^{\min }(f)$ to $\mathcal{B}_{\mathrm{ww}}^{\max }(f)$, and the Hamming weight of any bitstream $\Delta X^{1}\|\cdots\| \Delta X^{i}\left\|\Delta Y^{1}\right\| \cdots \| \Delta Y^{j}$ ranges from $\mathcal{B}_{\mathrm{bw}}^{\min }(f)$ to $\mathcal{B}_{\mathrm{bw}}^{\max }(f)$.

Example 2.8. Consider the function $f_{\oplus}: \mathbb{B}^{n} \times \mathbb{B}^{n} \rightarrow \mathbb{B}^{n}$ such that $f_{\oplus}\left(X^{1}, X^{2}\right)=X^{1} \oplus X^{2}$. We have: $\mathcal{B}_{\mathrm{ww}}^{\min }\left(f_{\oplus}\right)=\mathcal{B}_{\mathrm{ww}}^{\max }\left(f_{\oplus}\right)=\mathcal{B}_{\mathrm{bw}}^{\min }\left(f_{\oplus}\right)=2$ and $\mathcal{B}_{\mathrm{bw}}^{\max }\left(f_{\oplus}\right)=2 n$. From $\mathcal{B}_{\mathrm{ww}}^{\min }\left(f_{\oplus}\right)=2$, we can deduce that at least two of $\left(X^{1}, X^{2}, X^{1} \oplus X^{2}\right)$ are nonzero if some of $\left(X^{1}, X^{2}\right)$ is nonzero.

Similarly, $\mathcal{B}_{\mathrm{ww}}^{\min }\left(f_{\odot}\right), \mathcal{B}_{\mathrm{ww}}^{\max }\left(f_{\odot}\right), \mathcal{B}_{\mathrm{bw}}^{\min }\left(f_{\odot}\right)$ and $\mathcal{B}_{\mathrm{bw}}^{\max }\left(f_{\odot}\right)$ for each bit-wise operation $\odot \in\{+,-, \wedge, \vee\}$ can be defined, whose values are given in Table 2 , where for clarity, we denote by $\mathcal{B}_{\mathrm{ww}, \odot}^{\min }, \mathcal{B}_{\mathrm{ww}, \odot}^{\max }, \mathcal{B}_{\mathrm{bw}, \odot}^{\min }$ and $\mathcal{B}_{\mathrm{bw}, \odot}^{\max }$ the corresponding numbers of the bit-wise operation $\odot \in\{+,-, \wedge, \vee, \oplus\}$. Furthermore, for a given matrix $M$ and S-box $\mathcal{S}, \mathcal{B}_{\mathrm{ww}, M}^{\min }, \mathcal{B}_{\mathrm{ww}, M}^{\max }, \mathcal{B}_{\mathrm{bw}, M}^{\min }, \mathcal{B}_{\mathrm{bw}, M}^{\max }, \mathcal{B}_{\mathrm{ww}, S}^{\min }, \mathcal{B}_{\mathrm{ww}, S}^{\max }, \mathcal{B}_{\mathrm{bw}, S}^{\min }$ and $\mathcal{B}_{\mathrm{bw}, S}^{\max }$ are defined accordingly for the linear transformation $f_{M}(x)=M * x$ and substitution $f_{\mathcal{S}}=\mathcal{S}(x)$.

## 3 THE DESIGN OF EASYBC

EAsyBC is a high-level, statically-typed, C-like cryptography-specific language, designed for conveniently describing block ciphers but without complicating the subsequent automated security analysis, so that cryptographers can quickly implement and analyze a block cipher especially during the design phase.

### 3.1 Syntax

The syntax of EAsyBC is given in Figure 1.
Boxes. EASYBC features one standard array type and four cryptography-specific array types decorated by sbox, pbox, pbox ${ }_{m}$ and ffm, respectively, that are commonly used for implementing block ciphers. The cryptography-specific arrays are global and immutable, thus called boxes in this paper. In particular, sbox defines an array that acts as a lookup-table based S-box for transforming $m$ input bits to $n$ output bits; pbox defines an array for performing permutation; pbox $x_{m}$ defines a matrix for linear transformation via matrix-vector product. (Note that pbox can be implemented via pbox ${ }_{m}$, we provide both for convenience.) ffm describes the finite-field multiplication $(\otimes)$ which may vary in block ciphers, thus should be defined by users. An ffm box is required only for performing

| Operation $\odot \in\{+,-, \wedge, \vee, \oplus\}$ | STMT $S::=\tau x$ | Declaration |
| :---: | :---: | :---: |
| $\mathrm{Operation}_{2} \star \in\{,-, *, /, \%\}$ |  | Assignment |
| Width $s::=1\|4\| 6\|8\| 16 \mid \cdots$ | $\mid x[\xi]=e$; | Array put |
| Constant $n::=0\|1\| 2\|3\| \cdots$ | $\mid x=f\left(e_{1}, \cdots, e_{n}\right)$; | Function call |
| Base Type $\tau::=$ uints \| uints[n]|uint | \| $\tau x=e$; | Decl-Init |
| Position $\xi::=n\|x\| \xi \star \xi$ | \| for ( $x$ from $n_{1}$ to $\left.n_{2}\right)\left\{S^{+}\right\}$ | Range-for |
| $\begin{array}{r} \text { Expression } e::=n\|x\| e_{1} \odot e_{2}\|\sim e\| M * e \\ \\ \|x\langle\cdot e \cdot\rangle\| x\langle e\rangle \mid \operatorname{View}\left(e, \xi_{1}, \xi_{2}\right) \end{array}$ | for ( $x$ from $n_{1}$ to $n_{2}$ ) $\left\{S^{+}\right\}$ | Range for |
| $\mid$ touint $\left(e_{0}, \cdots, e_{s-1}\right) \mid$ tou | $(e)\|e \lll \xi\| e \ggg \xi \mid e[\xi]$ |  |

Round_FN rnd::= r_fn uints[n] $f$ (uint $r$, uints[ $\left.n_{1}\right] s k$, uints[ $\left.n\right] p$ ) $\left\{S^{+}\right.$return $\left.y ;\right\}$
Sbox_FN sbox::= s_fn uints $s_{1} f\left(\right.$ uints $\left._{2} x\right)\left\{S^{+}\right.$return $\left.y ;\right\}$
Box box::= uints[ $n] x=\left\{n_{0}, \cdots, n_{m}\right\}$
$\mid$ sbox uints $[n] x=\left\{n_{0}, \cdots, n_{m}\right\} \mid \operatorname{pbox}$ uint $[n] x=\left\{n_{0}, \cdots, n_{m}\right\}$
| $\operatorname{pbox}_{m}$ uints $[n][n] M=\left\{\left\{n_{0,0}, \cdots, n_{0, m}\right\}, \cdots,\left\{n_{m, 0}, \cdots, n_{m, m}\right\}\right\}$
| ffmuints $[n][n] M=\left\{\left\{n_{0,0}, \cdots, n_{0, m}\right\}, \cdots,\left\{n_{t, 0}, \cdots, n_{t, m}\right\}\right\}$
Def defs::= box|sbox|rnd|defs defs
Program $\quad \mathrm{P}::=$ @cipher name defs fnuints[ $n$ ] $f\left(\right.$ uints $\left[n_{1}\right] k$, uints $\left.[n] p\right)\left\{S^{+}\right.$return $\left.y ;\right\}$

Fig. 1. Syntax of EasyBC.
linear transformations using pbox $\mathrm{x}_{\mathrm{m}}$, i.e., evaluating expressions of the form $M * x$, thus cannot be explicitly involved in any statements. In this work, we assume a finite field of characteristic 2 , so the finite-field addition is the bit-wise $\operatorname{XOR}(\oplus)$.
Positions. Position $\xi$ is used to express array indices for array get $(e[\xi])$, array slice $\left(\operatorname{View}\left(e, \xi_{1}, \xi_{2}\right)\right)$, array put $(x[\xi]=e)$ and array left/right-rotation ( $e \lll \xi$ and $e \gg \xi$ ) via the common operations $\{+,-, *, /, \%\}$, whose values can be statically determined after preprocessing (i.e., independent of inputs). After preprocessing, all the positions $\xi$ will be constants.
Expressions. Expressions are defined as usual, including modular addition (+), modular substitution $(-)$, bit-wise AND ( $\wedge)$, bit-wise OR ( $\vee$ ), bit-wise NOT $(\sim)$ and bit-wise $\operatorname{XOR}(\oplus)$, as well as common cryptography-specific operations.
$M * e$ is a matrix-vector product using finite-field multiplication $(\otimes)$ and addition $(\oplus)$, where the matrix $M$ must be defined as an array of type pbox $_{m}$ uints $[n][n]$ and the vector $e$ should be an array of type uints $[n] . x\langle\cdot e\rangle$ is provided for performing permutation, where $x$ is a P-box. Similarly, $x\langle e\rangle$ is provided for performing substitution, where $x$ is an $\operatorname{S-box}$. View $\left(e, \xi_{1}, \xi_{2}\right)$ is a slice of the array $e$ starting at the index $\xi_{1}$ and ending at the index $\xi_{2}$ (inclusive), e.g., View ( $\left.(a, b, c, d), 1,3\right)$ is ( $b, c, d$ ). touint $\left(e_{0}, \cdots, e_{s-1}\right)$ transforms the $s$-bitstream $\left(e_{0}, \cdots, e_{s-1}\right)$ into an $s$-bit unsigned integer $x$ such that $\operatorname{bin}(x)=\left(e_{0}, \cdots, e_{s-1}\right)$, e.g., touint $(1,1,0,1)$ is the 4 -bit unsigned integer 13 . touint $(e)$ is the same as touint $(e[0], \cdots, e[n-1])$ for the array $e$ of type uint1[n]. An array $e$ can be left (resp. right) rotated $\xi$ positions via $e \lll \xi$ (resp. $e \gg \xi$ ), which can be seen special length-preserving permutations. $e[\xi]$ is array get, the same as $\operatorname{View}(e, \xi, \xi)$.
Statements. Statements in EASYBC can be declarations, assignments, array puts, returns and round function calls. Note that EAsYBC does not support branching statements (e.g., if-then-else), because current EAsYBC suffices to implement both encryption and decryption processes of block ciphers while branching statements will make modeling complicated when analyzing security. We will evaluate the expressive capability of EAsYBC in Section 8.1.

Statement $\tau x=e$ is a syntactic sugar of $\tau x ; x=e$. Statement for $\left(x\right.$ from $n_{1}$ to $\left.n_{2}\right)\{S\}$ is a range-for loop. Note that the range $\left[n_{1}, n_{2}\right.$ ] is limited to constants, which suffices to express block ciphers. The range variable $x$ should be typed as uint, thus could be used in computing indices $\xi$.

Functions. EAsyBC has three types of functions decorated by $r_{-} f n, s_{-} f n$ or $f n$. A function decorated by $r_{-} f n$ is a round function, whose formal parameters are fixed to be the round number $r$, subkey $s k$, and text $t x t$. Note that the subkey $s k$ and input text $t x t$ should have the same element type uints but may differ in number of elements, and the input text $t x t$ and output of a round function should have the same type uints[ $n$ ]. An s_fn function is an alternative way to perform substitution instead of using arrays. Small S-boxes (e.g., 4-bit S-box in PRESENT [Bogdanov et al. 2007]) can be easily expressed as arrays, which can facilitate the follow-up security analysis. However, it would be infeasible to express large S-boxes as arrays (e.g., 64-bit S-box [Beierle et al. 2020]), for which s_fn functions can be used. An fn function defines the encryption (resp. decryption) process of a block cipher. Its parameters include the key $k$ and plaintext (resp. ciphertext) txt. The function body comprises declarations and round function calls for computing the ciphertext (resp. plaintext).
Programs. A program $P$ in EAsyBC consists of a cipher name (decorated by @cipher), definitions of global boxes and functions, and a definition of an fn function describing the cipher.

In this work, following [Mouha et al. 2011; Wu and Wang 2012; Zhang et al. 2018; Zhou et al. 2019], we consider single-key differential cryptanalysis, namely, the key is fixed and has no difference in any pair of executions, thus key schedule algorithms are omitted in EAsyBC programs. Nevertheless, EAsyBC can be easily extended to related-key differential cryptanalysis [Biryukov and Nikolić 2010] where the key may differ in some pairs of executions, by introducing a key schedule function. We leave this as interesting future work.
Core language. The (full) language of EasyBC is designed for conveniently describing block ciphers with rich but redundant constructs. To ease the automation of the subsequent security analysis, we have identified a subset of EasyBC as the core language (cf. Figure 1), i.e., in the places highlighted in yellow, positions $\xi$ and expressions $e$ are limited to constants and variables, respectively; in the places highlighted in grey constructs will not be present (i.e., they are to be eliminated by preprocessing the program written in the full language). A program in the core language is obtained by performing loop unrolling, constant-folding, constant propagation and dead-code elimination, on which type checking and security analysis are performed.

Example 3.1. Figure 2 shows a snippet of the 64 -bit block cipher PRESENT in EAsyBC. The array $s$ is an $S$-box for substitution, the array $p$ is the P-box for permutation. The round function $f 1$ is invoked during the 1 -st to the 31 -st round. Given the subkey sk and text t , $\mathrm{f} 1(i, \mathrm{sk}, \mathrm{t})$ for $1 \leq i \leq 31$ produces the output $r$ tn of $i$-th round. In detail, the input $t$ is XORed with the subkey sk and results in the array $n t$, then the second range-for loop slices nt into 16 arrays via calling View, each of which is substituted via the S-box s, resulting in the array s_out. The array s_out is processed by applying the array $p$ to perform permutation. Finally, the result $r t n$ returned.

Remarks on the design choice of EASYBC. In EASYBC, we introduce high-level constructs View, touint and permutations (i.e., $x(\cdot e \cdot)$ and $M * e$ ) to ease the implementation of block ciphers. The inputs and outputs of round functions are fixed-size blocks which can be implemented by arrays (e.g., t and rtn in Figure 2). Typically, a block is to be split into small ones on which a look-up table based S-box (e.g., array in EASYBC) of suitable size is applied (e.g., 4-bit S-box s in Figure 2). The outputs of S-boxes will be juxtaposed to form a large block on which permutations are performed via either matrix-vector product or P-boxes (e.g., p1 <-s_out.) in Figure 2). On the other hand, to ease the automation of the subsequent security analysis, most indices are limited to positions $\xi$ whose values can be statically determined (e.g., e[ $\xi]$ and $\operatorname{View}\left(e, \xi_{1}, \xi_{2}\right)$ ), and thus become constants after preprocessing. The loop-up table based S-boxes (i.e., $x\langle e\rangle$ ) are an exception as they require a

```
@cipher PRESENT
sbox uint4[16] s}={12,5,6,11,9,0,10,13,3,14,15,8,4,7,1,2}
pbox uint[64] p = {0,16,32,48,1,17,33,\ldots.15,31,47,63};
r_fn uint1[64] f1(uint r, uint1[64] sk, uint1[64] t){
    uint1[64] nt;
    for(i from 0 to 63){nt[i] = t[i]^ sk[i]; }
    uint1[64] s_out;
    for(i from 0 to 15) {
        uint1[4] temp = View(nt, i*4, i * 4 + 3);
        uint4 sbox_in = touint(temp[0], temp[1], temp[2], temp[3]);
        uint4 sbox_out = s1\langlesbox_in\rangle # substitution via S-box
        s_out[i*4]=sbox_out[0]; s_out[i*4+1]= sbox_out[1];
        s_out[i*4+2]=sbox_out[2]; s_out[i*4+3]= sbox_out [3];
    }
    uint1[64] rtn = p1\langle·s_out \rangle; # permutation via P-box
    return rtn; }
fn uint1[64] enc(uint1[2048] key, uint1[64] plaintext){
    uint1[64] text= plaintext;
    for(i from 1 to 31){ text=f1(i,View(key,(i - 1)*64,i*64-1),text); }
    ... # execute the last rounds
    return text; }
```

Fig. 2. Code snippet of the 64 -bit block cipher PRESENT in EasyBC.
complicated modeling method. It remains open how to handle generic array access with variable indices although currently there appears no such need to specify block ciphers.

### 3.2 Operational Semantics

Let $\mathbb{X}$ denote a set of variables. An (evaluation) context $\sigma: \mathbb{X} \rightarrow \bigcup_{i \geq 1} \mathbb{N}^{i}$ is a mapping from variables to values, where a value can be a (fixed-width) non-negative integer or an array. Let $\sigma[x \mapsto v]$ be the context such that $\sigma[x \mapsto v](y)=v$ if $x=y$, otherwise $\sigma[x \mapsto v](y)=\sigma(y)$.

The evaluation judgement is in the form of

$$
\sigma \vDash e: v
$$

meaning that the expression $e$ evaluates to the value $v$ under the evaluation context $\sigma$.
The evaluation rules are given in Figure 3 (top-part), most of which are standard. Rule (S-Box) states that $x\langle y\rangle$ is a substitution, namely, the entry of the S-box $\sigma(x)$ at the index $\sigma(y)$. Rule ( $\mathrm{P}-\mathrm{Box}_{1}$ ) states that $M * x$ is the matrix-vector product of the matrix $M$ and the array $\sigma(x)$. Rule ( $\mathrm{P}_{-\mathrm{Box}_{2} \text { ) }}$ ) states that $x\langle\cdot y \cdot\rangle$ is a permutation of the array $\sigma(y)$ according to the indices given by the P-box $\sigma(x)=\left(j_{0}, \cdots, j_{n-1}\right)$, where its entry at the index $i$ is the entry of the array $\sigma(y)$ at the index $j_{i}$.

The operational semantics of statements is defined as transition rules of the form

$$
(\sigma, S) \Rightarrow \sigma^{\prime}
$$

meaning that the execution of the statement $S$ from the state $\sigma$ results in the state $\sigma^{\prime}$. For a sequence of statements $S_{1} ; S_{2} ; \cdots S_{n} ;\left(\sigma_{0}, S_{1} ; S_{2} ; \cdots S_{n} ;\right) \Rightarrow^{+} \sigma_{n}$ denotes the transitive transition of $\Rightarrow$, i.e., $\left(\sigma_{0}, S_{1}\right) \Rightarrow \sigma_{1},\left(\sigma_{1}, S_{2}\right) \Rightarrow \sigma_{2}, \cdots,\left(\sigma_{n-1}, S_{n}\right) \Rightarrow \sigma_{n}$. The transition rules of EAsYBC are listed in Figure 3 (bottom-part), which are standard. We denote by $\sigma_{0} S_{1} \sigma_{1} S_{2} \sigma_{2} \cdots S_{n} \sigma_{n}$ the execution of the program $P$ starting from the state $\sigma_{0}$ and ending at the state $\sigma_{n}$, and ( $\left.\sigma_{i-1}, S_{i}\right) \Rightarrow \sigma_{i}$ for $1 \leq i \leq n$.

For an execution of an $r$-IBC, we have
(1) round functions can only be invoked in the fn function,
(2) the first arguments in the invoked round functions are the round numbers $1,2,3, \cdots, r$, and

Fig. 3. The operational semantics of core EAsyBC, where $\odot \in\{+,-, \oplus, \wedge, \vee\}$.
(3) the input and output of $i$-th round are the third argument and return value of the invoked round function whose first argument is $i$.

### 3.3 Type System of Core EasyBC

The type system of core EASyBC is designed to disallow certain kinds of illegal programs and provide type information for security analysis.

EAsyBC supports the following types, i.e.,

$$
\beta::=\text { uints | uint | uints[n]| sbox uints[n]| pbox uint }[n] \mid \operatorname{pbox}_{m} \text { uints }[n][n] .
$$

Here, uints is for $s$-bit unsigned integers, uints[ $n$ ] is for vectors (or arrays) of $s$-bit unsigned integers, uint is for unsigned integers, and uint $[n]$ is for vectors (or arrays) of unsigned integers. Note that uints is identical to uint1[s] and uints[1]. uint and uint [ $n$ ] are used only when the variables under typing are independent of inputs.
Typing expressions. The typing judgement is of the form of

$$
T_{g}, T_{l} \vdash e: \beta
$$

where $\left(T_{g}, T_{l}\right)$ is a typing context, $e$ is an expression under typing, and $\beta$ is a type. The global environment $T_{g}$ is a mapping from global variables to their types and from function names to function signatures $\left(\beta_{0}, \cdots, \beta_{n}\right)$ where $\beta_{0}$ is the return type and $\beta_{1}, \cdots, \beta_{n}$ are the types of the formal parameters. The local environment $T_{l}$ is a mapping from local variables to their types. The typing judgement $T_{g}, T_{l} \vdash e: \beta$ is valid if $e$ has type $\beta$ under the typing context $\left(T_{g}, T_{l}\right)$.

Figure 4 (top-part) gives typing rules for expressions. Rule (T-Uints) express that a non-negative integer $n$ can be typed as uints if $n \leq 2^{s}-1$. Rules (T-VAR) and (T-Not) are defined as usual. Rule (T-Op) ensures that the two operands and result of the operation $\odot \in\{+,-, \oplus, \wedge, \vee\}$ have the same type uints. Rule (T-Pbox ${ }_{1}$ ) ensures that the array $x$ has suitable type w.r.t. the type of the matrix $M$ for matrix-vector product. Rule (T- $\mathrm{PBOx}_{2}$ ) requires that the elements in the array $x\langle\cdot y \cdot\rangle$ and the

Fig. 4. The typing rules of core EasyBC.
operand $y$ have the same type, as the P-box $x$ only specifies the element order for the permutation. Rule (T-Sbox) requires that the S-box has a sufficient number of elements (i.e., $n \geq 2^{s_{2}}$ ) and the result $x\langle y\rangle$ has the same type as the elements in the S-box $x$. Note that both S-boxes and P-boxes do not necessarily preserve the length which may occur, e.g., DES. Rule (T-View) requires that the indices $n_{1}$ and $n_{2}$ are within the bounds of the array $x$, moreover, the slice $\operatorname{View}\left(x, n_{1}, n_{2}\right)$ is an array with length $n_{2}-n_{1}+1$ and its elements have the same type as the elements in the array $x$. Rule (T-Toint) requires that all the operands $x_{i}$ have type uint1 and the result has type uints where $s$ is the number of operands.
Typing statements. The typing judgement of a statement is in the form of

$$
T_{g}, T_{l}, f \vdash S
$$

where ( $T_{g}, T_{l}$ ) is a typing context, $S$ is the statement under typing in the function $f$. We write $T_{g}, T_{l}, f \vdash S$ is valid if $S$ is well-typed.

The typing rules are given in Figure 4 (middle-part). Rule (T-Dect) is defined as usual. Rule (T-Ass) requires that the type of the expression $e$ conforms to the declared type of the variable $x$. Rule (T-Arr-Put) requires that the index $i$ is within the bounds of the array $x$ and the operand $y$ has the same type as the elements in the array $x$. Rule (T-CALL) requires that the types of actual arguments and return conform to the corresponding function signature $T_{g}\left(f^{\prime}\right)$.
Typing programs. Each program $P$ is typed by iteratively typing each function definition. The program is well-typed if all the function definitions are well-typed. The typing judgement of a function definition fn _def is in the form of

$$
T_{g}, T_{l} \vdash f n_{\_} \text {def },
$$



Fig. 5. Overview of our approach.
where ( $T_{g}, T_{l}$ ) is a typing context and fn_def is a function definition under typing. The typing judgement $T_{g}, T_{l} \vdash \mathrm{fn} \_$def is valid if the function definition fn _def is well-typed. The typing rule (T-Fn-Def) is given in Figure 4 (bottom-part), which enforces the well-typed function body when the formal parameters and return have declared types.

### 3.4 Compilation

We have implemented an interpreter in C++ for EAsYBC to test the operational semantics of programs, making sure that they are consistent with the execution of reference block ciphers. In particular, we compare the output of EASYBC programs with that of running the binary executable compiled from C/C++ programs by GNU C++ compiler (G++). For each cryptographic primitive, we randomly generate inputs and then run the EasyBC program (with our interpreter) and the binary executable. We record their output, as well as the execution time for analysis. The results are given in Section 8.1.

### 3.5 Overview of Analysis

Recall that we are interested in evaluating the resistance of block ciphers against differential cryptanalysis by bounding the MaxEDCP. A block cipher is considered to be resistant to differential cryptanalysis if MaxEDCP is no greater than $O\left(2^{6}\right)$ for the block size $a$.

Figure 5 gives an overview of our approach. Given a program in full EAsyBC together with an option for selecting a particular MILP generation approach and an S-box modeling technique, EAsyBC computes an upper bound of the MaxEDCP. The result is conclusive if this upper bound is sufficient to show the resistance of the program. Our approach is not necessarily complete (e.g., in most cases we only compute an upper bound of MaxEDCP), so it may fail to prove the resistance of some programs, although this does not happen in our evaluation (cf. Section 8).

First, the input program is preprocessed to eliminate range-for loops and positional variables by performing loop unrolling, constant-folding, constant propagation and dead-code elimination. The final program will be in the core language of EasyBC. Hereafter, we assume that the given EasyBC program has been preprocessed.

Next, the program is type-checked to disallow certain kinds of illegal programs, e.g., the types of operands in expressions, formal parameters in function definitions and actual arguments in function calls are proper. It also provides type information for security analysis, in particular, the lengths of arrays, the type and the bit widths of array elements, which are used for MILP generation.

After type-checking, we reduce the problem of bounding the MaxEDCP to MILP. The key insight of the reduction is to characterize the dependency (i.e., feasibility) between input and output differences of each operation using integer linear (IL) constraints and bound the MaxEDCP by minimizing an objective function subject to the IL constraints. By utilizing an MILP solver (e.g., Gurobi [Gurobi Optimization 2018]), we can obtain an upper bound of the MaxEDCP. In practice, one may be only interested in proving the resistance against differential cryptanalysis. Hence we
also verify whether the MaxEDCP is no greater than a given threshold, an affirmative answer to which would be sufficient to show that the given cipher is resistant against differential cryptanalysis. This strategy is very effective in practice, as few rounds are often sufficient to prove the resistance by leveraging the decomposition approach (cf. Proposition 5.3 and Proposition 7.3).

To this end, we present two different approaches for reducing to MILP. The first one is by determining the lower bound of the minimum number $\mathcal{N}_{\text {diff }}$ of active S-boxes in either wordwise (Section 5) or bit-wise (Section 6) manner, because the MaxEDCP of $s$-round differential characteristics is bounded from above by $p^{N_{\text {diff }}}$ [Heys 2002; Sun et al. 2014a], where $p$ denotes the maximum probability $\operatorname{Pr}_{\mathcal{S}}(\Delta X, \Delta Y)$ among all the nonzero differentials $(\Delta X, \Delta Y)$ for any active $S$-box. Intuitively, the word-wise one models the difference of an $s$-bitstream under two executions by only one Boolean variable, thus is less involved and produces fewer constraints, but is limited to certain block ciphers e.g., it cannot be directly applied to bit-oriented block ciphers such as PRESENT). In contrast, the bit-wise approach models the difference of each bit by one Boolean variable, thus is more fine-grained and has wider applicability. One may understand that the bitwise approach implicitly bit-blasts the program and then generates MILP similar to the word-wise approach. The MILP generation in this approach requires the (maximum/minimum word-/bit-wise) branch numbers of some operations and representing S-boxes as IL constraints, for which we propose novel SMT-based methods to automatically determine branch numbers for each operation and implement some recent promising bit-wise S-box modeling techniques.

The first approach is efficient and often effective, but the obtained upper bound may not be sufficiently tight and is not applicable for some ciphers. We provide an extended bit-wise approach to directly bound the MaxEDCP (Section 7). In the extended bit-wise approach, the probabilities between input and output differences for each operation are further encoded into IL constraints using additional Boolean variables. This approach may be less efficient but is more accurate than the first approach. We also propose a novel Maximal Satisfiability Modulo Theories (MaxSMT) [Bjørner and Phan 2014] based extended bit-wise S-box modeling method which guarantees that the least number of Boolean variables is used for encoding differential probabilities of S-boxes.

## 4 UTILITIES

In this section, we present the three key utilities used in our MILP generation.

### 4.1 SMT-based Method for Determining Branch Numbers

The branch numbers of some operations are required in MILP generation. However, to our best knowledge, existing work usually relies on manual analysis. In this paper, we propose to determine branch numbers of a given operation/function, by reducing to the optimization problem modulo bit-vector theory, which can be solved by off-the-shelf optimizing SMT solver, e.g., Z3 [Bjørner et al. 2015].

Given a function $f: \mathbb{B}^{n_{1}} \times \cdots \times \mathbb{B}^{n_{i}} \rightarrow \mathbb{B}^{m_{1}} \times \cdots \times \mathbb{B}^{m_{j}}$, to compute its minimum (resp. maximum) word-wise branch number $\mathcal{B}_{\mathrm{ww}}^{\min }(f)$ (resp. $\mathcal{B}_{\mathrm{ww}}^{\max }(f)$ ), by Definition 2.7, we express the condition $\operatorname{BNCond}(f)$ and the additional condition $d=\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ as a quantifier-free SMT formula $\phi_{f}$ in the bit-vector theory, and use Z 3 to minimize (resp. maximize) the variable $d$ subject to the SMT formula $\phi_{f}$. The optimized value of $d$ is $\mathcal{B}_{w w}^{\min }(f)$ (resp. $\mathcal{B}_{w w}^{\max }(f)$ ). An illustrating example is given in [Sun et al. 2023, Section C.1].

The minimum (resp. maximum) bit-wise branch number $\mathcal{B}_{\mathrm{bw}}^{\min }(f)$ (resp. $\mathcal{B}_{\mathrm{bw}}^{\max }(f)$ ) can be computed the same as above, except that $\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ counts the number of 1 bits in the bitstream $\Delta X^{1}\|\cdots\| \Delta X^{i}\left\|\Delta Y^{1}\right\| \cdots \| \Delta Y^{j}$.

Proposition 4.1. The minimized (maximized) value of $d$ is

- $\mathcal{B}_{w w}^{\min }(f)\left(\right.$ resp. $\left.\mathcal{B}_{w w}^{\max }(f)\right)$ when $\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ counts the number of nonzero entries in the vector $\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$,
- $\mathcal{B}_{\mathrm{bw}}^{\min }(f)\left(\right.$ resp. $\left.\mathcal{B}_{\mathrm{bw}}^{\max }(f)\right)$ when $\operatorname{cnt}\left(\Delta X^{1}, \cdots, \Delta X^{i}, \Delta Y^{1}, \cdots, \Delta Y^{j}\right)$ counts the number of 1 bits in the bitstream $\Delta X^{1}\|\cdots\| \Delta X^{i}\left\|\Delta Y^{1}\right\| \cdots \| \Delta Y^{j}$.


### 4.2 Bit-wise S-box Modeling

Consider an expression $\mathcal{S}\langle X\rangle$ where $\mathcal{S}: \mathbb{B}^{n} \rightarrow \mathbb{B}^{m}$ is a lookup-table based S-box. It is non-trivial to specify the bit-level dependency of differences between the input $X$ and the result of $\mathcal{S}\langle X\rangle$ (denoted by $Y$ ), as the S-box provides only input and output pairs. To resolve this issue, we first compute its DDT $\mathcal{D}_{\mathcal{S}}: \mathbb{B}^{n} \times \mathbb{B}^{m} \rightarrow \mathbb{N}$ from which IL constraints are generated to characterize the bit-level dependency of differences between $X$ and $Y$. Recall that for every differential $(\Delta X, \Delta Y) \in \mathbb{B}^{n} \times \mathbb{B}^{m}$, $\mathcal{D}_{\mathcal{S}}(\Delta X, \Delta Y)$ gives the number of inputs $X \in \mathbb{B}^{n}$ such that $\mathcal{S}(X) \oplus \mathcal{S}(X \oplus \Delta X)=\Delta Y$.

Let $\vec{b}$ be the input difference $\Delta X$ and $\vec{b}^{\prime}$ be the output difference $\Delta Y$ of the S-box $\mathcal{S}$. We have

- $\sum_{i=0}^{n-1} \vec{b}_{i}=0 \Rightarrow \sum_{i=0}^{m-1} \vec{b}_{i}^{\prime}=0$, i.e., no output difference if no input difference, namely, if two inputs to $\mathcal{S}$ are the same, then the two outputs are the same. This condition can be characterized by the IL constraint $m \cdot \sum_{i=0}^{n-1} \vec{b}_{i} \geq \sum_{i=0}^{m-1} \vec{b}_{i}^{\prime}$, denoted by $\Psi_{\mathcal{S}}^{1}$.
- $\sum_{i=0}^{m-1} \vec{b}_{i}^{\prime}=0 \Rightarrow \sum_{i=0}^{n-1} \vec{b}_{i}=0$ if $\mathcal{S}$ is injective, i.e., no input difference if no output difference, namely, if two outputs of $\mathcal{S}$ are the same, then the two inputs must be the same. This condition can be exactly characterized by the IL constraint $n \cdot \sum_{i=0}^{m-1} \vec{b}_{i}^{\prime} \geq \sum_{i=0}^{n-1} \vec{b}_{i}$, denoted by $\Psi_{\mathcal{S}}^{2}$.
- $\mathcal{D}_{\mathcal{S}}\left(\vec{b}, \vec{b}^{\prime}\right) \neq 0$, namely, $\left(\vec{b}, \vec{b}^{\prime}\right)$ should be feasible for $\mathcal{S}$. We implement and compare the promising techniques [Abdelkhalek et al. 2017; Boura and Coggia 2020; Li and Sun 2022; Sasaki and Todo 2017; Sun et al. 2014b; Udovenko 2021] that can characterize $\mathcal{D}_{\mathcal{S}}\left(\vec{b}, \overrightarrow{b^{\prime}}\right) \neq 0$ by IL constraints. Hereafter, we denote by $\Psi_{\mathcal{S}}^{3}$ the set of IL constraints such that $\left(\vec{b}, \vec{b}^{\prime}\right)$ is a solution of $\Psi_{\mathcal{S}}^{3}$ iff $\mathcal{D}_{\mathcal{S}}\left(\vec{b}, \vec{b}^{\prime}\right) \neq 0$.
Proposition 4.2. $\left(\vec{b}, \vec{b}^{\prime}\right)$ is feasible input and output differences of $\mathcal{S}$ iff $\left(\vec{b}, \vec{b}^{\prime}\right)$ is a solution of $\Psi_{\mathcal{S}}^{3}$.
We denote by $\Psi_{\mathcal{S}}$ the set $\Psi_{\mathcal{S}}^{1} \cup \Psi_{\mathcal{S}}^{2} \cup \Psi_{\mathcal{S}}^{3}$ if the S-box $\mathcal{S}$ is injective, otherwise $\Psi_{\mathcal{S}}^{1} \cup \Psi_{\mathcal{S}}^{3}$.


### 4.3 MaxSMT-based Extended Bit-wise S-box Modeling

The above bit-wise S-box modeling method is able to characterize all the feasible differentials $(\Delta X, \Delta Y) \in \mathbb{B}^{n} \times \mathbb{B}^{m}$ of the $S$-box $\mathcal{S}$, but the probability $\operatorname{Pr} \mathcal{S}_{\mathcal{S}}(\Delta X, \Delta Y)=\frac{\mathcal{D}_{\mathcal{S}}(\Delta X, \Delta Y)}{2^{n}}$ of differentials is not present in the IL constraints, so it is impossible to bound the MaxEDCP directly. We propose a novel MaxSMT-based method which guarantees that the least number of Boolean variables is used for encoding probabilities of differentials.

Definition 4.3. Given two sets of constraints ( $\Phi_{1}, \Phi_{2}$ ), the MaxSMT problem is to find a solution that satisfies all the constraints in $\Phi_{1}$ and maximizes the number of satisfied constraints in $\Phi_{2}$.

Let $V=\left\{v_{1}, \cdots, v_{h}\right\}$ be the set of nonzero probabilities $\operatorname{Pr} \mathcal{S}(\Delta X, \Delta Y)$ for $(\Delta X, \Delta Y) \in \mathbb{B}^{n} \times \mathbb{B}^{m}$. We define the MaxSMT problem $\left(\Phi_{1}^{\mathcal{S}}, \Phi_{2}^{\mathcal{S}}\right)$ for the S-box $\mathcal{S}$, where

$$
\Phi_{1}^{\mathcal{S}}=\left\{\sum_{i=1}^{h} c_{i} \cdot p_{i, j}=-\log _{2} v_{j} \mid 1 \leq j \leq h\right\} \text { and } \Phi_{2}^{\mathcal{S}}=\left\{c_{1}=0, \cdots, c_{h}=0\right\},
$$

for $1 \leq i, j \leq h, c_{i}$ is a variable over real numbers and $p_{i, j}$ is a Boolean variable. Clearly, for every $1 \leq j \leq h, 2^{-\sum_{i=1}^{h} c_{i} \cdot p_{i, j}}$ retains the probability $v_{j}$. Since the Boolean variable $p_{i, j}$ can be treated as a variable over real numbers by adding $p_{i, j}=0 \vee p_{i, j}=1$ to $\Phi_{1}^{\mathcal{S}}$, and $v_{j}$ (hence $\log _{2} v_{j}$ ) is a constant for each $1 \leq j \leq h$, the problem ( $\Phi_{1}^{\mathcal{S}}, \Phi_{2}^{\mathcal{S}}$ ) is a MaxSMT problem (modulo the theory of real numbers).

A solution of the MaxSMT problem $\left(\Phi_{1}^{\mathcal{S}}, \Phi_{2}^{\mathcal{S}}\right)$ assigns values to the variables $c_{i}$ 's and $p_{i, j}$ 's from which the probability $v_{j}$ for every $1 \leq j \leq h$ can be obtained. Suppose the solution assigns the values $\left\{b_{1, j}, \cdots, b_{h, j}\right\}$ to the Boolean variables $\left\{p_{1, j}, \cdots, p_{h, j}\right\}$ and the values $\left\{t_{1}, \cdots, t_{h}\right\}$ to the variables $\left\{c_{1}, \cdots, c_{h}\right\}$, we have: $2^{-\sum_{i=1}^{h} t_{i} \cdot b_{i, j}}=v_{j}$. Note that $\left(\Phi_{1}^{\mathcal{S}}, \Phi_{2}^{\mathcal{S}}\right)$ is always satisfiable.

We can observe that if $t_{i}=0$, the value $b_{i, j}$ of the Boolean variable $p_{i, j}$ for $1 \leq j \leq h$ can be omitted for retaining all the probabilities in $V$. We will see later that the values $\left\{b_{1, j}, \cdots, b_{h, j}\right\}$ of the Boolean variables $p_{i, j}$ 's will be used to encode the probabilities in $V$, we define $\Phi_{2}^{\mathcal{S}}$ as $\left\{c_{1}=0, \cdots, c_{h}=0\right\}$ so that a solution of the MaxSMT problem $\left(\Phi_{1}^{\mathcal{S}}, \Phi_{2}^{\mathcal{S}}\right)$ maximizes the number of 0 bits in the values $\left\{t_{1}, \cdots, t_{h}\right\}$ of the variables $\left\{c_{1}, \cdots, c_{h}\right\}$, thus minimizing the number of additional Boolean variables used for encoding the probabilities in $V$.

Let $\left\{i_{1}, \cdots, i_{k}\right\}$ be the set of indices of the nonzero values in $\left\{t_{1}, \cdots, t_{h}\right\}$. To encode all the probabilities in $V$, we define an extended DDT $\mathcal{D}_{\mathcal{S}}^{\dagger}$ of the $S$-box $\mathcal{S}$ as follows:

$$
\forall(\Delta X, \Delta Y) \in \mathbb{B}^{n} \times \mathbb{B}^{m} .1 \leq j \leq h . \mathcal{D}_{\mathcal{S}}^{\dagger}\left(\Delta X, \Delta Y, b_{i_{1}, j}, \cdots, b_{i_{k}, j}\right) \neq 0 \text { iff } \operatorname{Pr} \mathcal{S}_{\mathcal{S}}(\Delta X, \Delta Y)=v_{j} .
$$

A set $\Psi_{\mathcal{S}}^{4}$ of constraints over the Boolean variables $\vec{b}, \overrightarrow{b^{\prime}}, p_{i_{1}}, \cdots, p_{i_{k}}$ can be generated from the extended DDT $\mathcal{D}_{\mathcal{S}}^{\dagger}$ (cf. Section 4.2) such that

$$
\mathcal{D}_{\mathcal{S}}^{\dagger}\left(\Delta X, \Delta Y, b_{i_{1}}, \cdots, b_{i_{k}}\right) \neq 0 \text { iff }\left(\Delta X, \Delta Y, b_{i_{1}}, \cdots, b_{i_{k}}\right) \text { is a solution of } \Psi_{\mathcal{S}}^{4} .
$$

Proposition 4.4. For any solution $\left(\Delta X, \Delta Y, b_{i_{1}}, \cdots, b_{i_{k}}\right)$ of $\Psi_{\mathcal{S}}^{4}, \operatorname{Pr}(\Delta X, \Delta Y)=2^{-\sum_{j=1}^{k} t_{i_{j}} \cdot b_{i_{j}}}$.
We denote by $\Psi_{\mathcal{S}}^{\dagger}$ the set $\Psi_{\mathcal{S}}^{1} \cup \Psi_{\mathcal{S}}^{2} \cup \Psi_{\mathcal{S}}^{4}$ if the S-box $\mathcal{S}$ is injective, otherwise $\Psi_{\mathcal{S}}^{1} \cup \Psi_{\mathcal{S}}^{4}$. An illustrating example is given in [Sun et al. 2023, Section C.2].

## 5 WORD-WISE APPROACH

In this section, we present an approach for determining the lower bound of the minimum number of active S-boxes by reducing to MILP in a word-wise fashion. It works for programs $P$ where types are uints $[n]$ for a fixed bit size $s$ of the involved S-boxes and individual bits of any entry in an array cannot be changed. Our tool can automatically check if the word-wise approach is applicable. Recall that uints, uint1[s] and uints[1] are identical, and we shall use uints[1] hereafter. Note that, in this setting, the program $P$ is free of touint expressions.
High-level intuition. Each entry of an array variable $x$ in the program $P$ is modeled as a Boolean variable $b$, where $b=1$ in the MILP solution indicates that the entry has a difference when $P$ is executed under two distinct inputs for some fixed key. Thus, a variable of type uints $[n]$ is modeled by a vector $\vec{b}$ of $n$ Boolean variables. Each statement is modeled by a set of IL constraints over the Boolean variables which characterize the propagation of differences through the statement. Furthermore, each S-box $\mathcal{S}$ is associated with a unique Boolean variable $b_{\mathcal{S}}$ such that the S -box $\mathcal{S}$ is active under two execution if $b_{\mathcal{S}}=1$ (cf. Definition 2.4). The objective function is to minimize the sum of Boolean variables $b_{\mathcal{S}}$ for all the S -boxes $\mathcal{S}$, subject to the extracted IL constraints. The MILP solution gives the lower bound of the minimum number of active S-boxes in the program.

The word-wise MILP generation rules are given by a word-wise differential denotational semantics of EAsyBC. We present the denotational semantics for expressions in Section 5.1 and the denotational semantics for statements in Section 5.2.

### 5.1 Word-wise Differential Denotational Semantics for Expressions

W denotes by $\mathbb{X}_{f}$ the set of its local variables of a function $f$, and by $\mathbb{K}_{f} \subseteq \mathbb{X}_{f}$ the set of variables that are subkeys. We denote by $|x|=n$ if $x$ has type uints [ $n$ ].

| $\llbracket \operatorname{View}(x, i, j) \rrbracket_{\gamma}^{\mathrm{W}}=\left(\emptyset,\left(\vec{b}_{i}, \cdots, \vec{b}_{j}\right)\right)$, where $\vec{b}=\gamma(x)$ | $\sim x \rrbracket_{\gamma}^{\mathrm{W}}=(\emptyset, \gamma(x))$ |
| :---: | :---: |
| $\llbracket x_{1}+x_{2} \rrbracket_{Y}^{\mathrm{W}}=\llbracket x_{1}-x_{2} \rrbracket_{Y}^{\mathrm{W}}=\llbracket x_{1} \oplus x_{2} \rrbracket_{Y}^{\mathrm{W}}=\left(\Psi_{2,3}^{i}\left(b_{0}, b_{1}, b_{2}\right), b_{0}\right)$, where $i=1,2, b_{0}=$ newBV() |  |
| $\Psi_{2,3}^{1}\left(b_{0}, b_{1}, b_{2}\right)=\left\{b_{1}+b_{2} \geq b_{0}, b_{0}+b_{1} \geq b_{2}, b_{0}+b_{2} \geq b_{1}\right\}$ [Li et al. 2019], $b_{1}=\gamma\left(x_{1}\right), b_{2}=\gamma\left(x_{2}\right)$ |  |
| $\Psi_{2,3}^{2}\left(b_{0}, b_{1}, b_{2}\right)=\left\{b^{\prime} \geq b_{0}, b^{\prime} \geq b_{1}, b^{\prime} \geq b_{2}, \sum_{i=0}^{2} b_{i} \geq 2 b^{\prime}\right\}$ [Mouha et al. 2011], $b^{\prime}=\operatorname{newBV}()$ |  |
| $\llbracket x_{1} \wedge x_{2} \rrbracket_{\gamma}^{\mathbb{W}}=\llbracket x_{1} \vee x_{2} \rrbracket_{\gamma}^{\mathbb{W}}=\left(\left\{b_{1}+b_{2} \geq b_{0}\right\}, b_{0}\right)$, where $b_{0}=\operatorname{newBV}(), b_{1}=\gamma\left(x_{1}\right), b_{2}=\gamma\left(x_{2}\right)$ |  |
| $\begin{aligned} & \\| M * x \rrbracket_{Y}^{W}=\left(\Psi_{M}^{i}\left(\vec{b}, \vec{b}^{\prime}\right), \vec{b}^{\prime}\right) \text {, where } i=1,2, \vec{b}=\gamma(x), \vec{b}^{\prime}=\text { newBV }(), b^{\prime \prime}=\text { newBV }() \\ & \Psi_{M}^{1}\left(\vec{b}, \vec{b}^{\prime}\right)=\left\{\mathcal{B}_{\left.\mathcal{W w n}_{w, M}^{\min } \cdot b^{\prime \prime} \leq \sum_{i=0}^{\|x\|-1}\left(\vec{b}_{i}+\vec{b}_{i}^{\prime}\right) \leq \mathcal{B}_{\mathrm{ww}}^{\max }, 2\|x\| \cdot b^{\prime \prime} \geq \sum_{i=0}^{\|x\|-1}\left(\vec{b}_{i}+\vec{b}_{i}^{\prime}\right)\right\}}^{\Psi_{M}^{2}\left(\vec{b}, \vec{b}^{\prime}\right)=\left\{\mathcal{B}_{\mathrm{ww}, M}^{\min } \cdot b^{\prime \prime} \leq \sum_{i=0}^{\|x\|-1}\left(\vec{b}_{i}+\vec{b}_{i}^{\prime}\right) \leq \mathcal{B}_{\mathrm{ww}, M^{\prime}}^{\max }, b^{\prime \prime} \geq \vec{b}_{0}, b^{\prime \prime} \geq \vec{b}_{0}^{\prime}, \cdots, b^{\prime \prime} \geq \vec{b}_{\|x\|-1}, b^{\prime \prime} \geq \vec{b}_{\|x\|-1}^{\prime}\right\}}\right\} \end{aligned}$ |  |
|  |  |
|  |  |
| $\llbracket x \leftharpoonup y \cdot\rangle \rrbracket_{\gamma}^{W}=\left(\emptyset,\left(\vec{b}_{j_{0}}, \cdots, \vec{b}_{j_{n-1}}\right)\right.$ ), where $x$ is P-box $\left(j_{0}, \cdots, j_{n-1}\right)$ and $\vec{b}=\gamma(y)$ |  |
| $\llbracket x\langle y\rangle \rrbracket_{\gamma}^{\mathbb{W}}=\left(\left\{b \diamond b^{\prime}\right\}, b^{\prime}\right)$, where $b=\gamma(y), b^{\prime}=$ new $\operatorname{BV}(), \diamond$ is $=$ if $x$ is injective, otherwise $\geq$ |  |

Fig. 6. The word-wise differential denotational semantic rules for expressions.

State. A state $\gamma$ is a mapping from array entries $(x, i) \in\left(\mathbb{X}_{f} \backslash \mathbb{K}_{f}\right) \times \mathbb{N}$ to Boolean variables that model the differences of array entries under two executions. For each variable $x \in \mathbb{X}_{f} \backslash \mathbb{K}_{f}$,

- $\gamma(x)$ gives the sequence of Boolean variables $\gamma(x, 0), \cdots, \gamma(x,|x|-1)$ of the array $x$.
- $\gamma[(x, i) \mapsto b]$ denotes the update of $\gamma$ by mapping $(x, i)$ to the Boolean variable $b$, and $\gamma[x \mapsto \vec{b}]$ denotes the update $\gamma\left[(x, 0) \mapsto \vec{b}_{0}\right] \cdots\left[(x,|x|-1) \mapsto \vec{b}_{|x|-1}\right]$ for a Boolean vector $\vec{b}$ with $|\vec{b}|=|x|$.
By abuse of notation, $\gamma(x)$ gives the vector $\overrightarrow{0}$ with $|\overrightarrow{0}|=|x|$ if $x$ is a constant or subkey variable in $\mathbb{K}_{f}$. Note that $\gamma(x)$ is $\gamma(x, 0)$ if $x$ has type uints[1]. We denote by $\Gamma$ the set of states.
Denotational semantics. The (word-wise differential) denotational semantics of an expression $e$ is given by $\llbracket e \rrbracket^{\mathrm{W}}$ that maps each state $\gamma \in \Gamma$ to a pair $(\Psi, \vec{b})$, denoted by $\llbracket e \rrbracket_{\gamma}^{\mathrm{W}}$, where
- $\Psi$ is a set of IL constraints over Boolean variables characterizing the dependency/feasiblity of the differences between the support variables and result of $e$ such that the differences is a solution of $\Psi$ iff these differences are feasible for support variables and result of $e$;
- $\vec{b}$ such that $|\vec{b}|=|e|$ is a vector of the Boolean variables/values that models the difference of the result of $e$ under two executions ( $\vec{b}$ may be written as $b$ if $|\vec{b}|=1$ and $\vec{b}_{0}=b$ ).

Denotational semantic rules. The semantics for expressions in EasyBC is shown in Figure 6, where the function new BV() returns a fresh Boolean variable $b$ or a vector $\vec{b}$ of fresh Boolean variables according to the context. The semantic rules $\llbracket \sim x \rrbracket_{\gamma}^{W}, \llbracket \operatorname{View}(x, i, j) \rrbracket_{\gamma}^{W}$ and $\llbracket x\langle\cdot y \cdot\rangle \rrbracket_{\gamma}^{W}$ are straightforward according to their operational semantics. We explain the others below.

- $\llbracket x_{1} \odot x_{2} \rrbracket_{\gamma}^{W}$ for $\odot \in\{+,-, \oplus, \wedge, \vee\}$ gives a pair $\left(\Psi\left(b_{0}, b_{1}, b_{2}\right), b_{0}\right)$, where $\Psi\left(b_{0}, b_{1}, b_{2}\right)$ characterizes the dependency of the differences $b_{0}, b_{1}$ and $b_{2}$ between $x_{1} \odot x_{2}, x_{1}$ and $x_{2}$ according to the maximum and minimum word-wise branch numbers of $\odot$ (cf. Definition 2.7), and $b_{0}=0$ if $b_{1}=b_{2}=0$. For instance, $\mathcal{B}_{\mathrm{WN},+}^{\min }=2$ and $\mathcal{B}_{\mathrm{wW},+}^{\max }=3$ (cf. Table 2 ), meaning that either $b_{0}=b_{1}=b_{2}=0$ or at least two of them are 1 (i.e., $2 \leq b_{0}+b_{1}+b_{2} \leq 3$ ). For easy reference, these IL constraints are denoted by $\Psi_{2,3}^{1}$ or $\Psi_{2,3}^{2}$. Note that the auxiliary Boolean variable $b^{\prime}$ in $\Psi_{2,3}^{2}$ is 0 iff $b_{0}+b_{1}+b_{2}=0$. - $\llbracket M * x \rrbracket_{\gamma}^{W}$ gives the pair $\left(\Psi_{M}^{i}\left(\vec{b}, \overrightarrow{b^{\prime}}\right), \vec{b}^{\prime}\right)$, where $\Psi_{M}^{i}\left(\vec{b}, \overrightarrow{b^{\prime}}\right)$ characterizes the dependency of the differences $\vec{b}$ and $\vec{b}^{\prime}$ between the entries in the arrays $x$ and $M \odot x$ according to the maximum and minimum word-wise branch numbers of the linear transformation $M \odot x$. Indeed, $\Psi_{M}^{i}\left(\vec{b}, \overrightarrow{b^{\prime}}\right)$ enforces that the sum of their input and output differences $\sum_{i=0}^{|x|-1}\left(\vec{b}_{i}+\vec{b}_{i}^{\prime}\right)$ either ranges from


Fig. 7. The word-wise differential denotational semantic rules for statements.
$\mathcal{B}_{\mathrm{ww}, M}^{\min }$ to $\mathcal{B}_{\mathrm{ww}, M}^{\max }$ or is 0 in two alternative ways $\Psi_{M}^{1}\left(\vec{b}, \overrightarrow{b^{\prime}}\right)$ and $\Psi_{M}^{2}\left(\vec{b}, \overrightarrow{b^{\prime}}\right)$, where the auxiliary Boolean variable $b^{\prime \prime}$ in an MILP solution is 0 iff $\sum_{i=0}^{|x|-1}\left(\vec{b}_{i}+\vec{b}_{i}^{\prime}\right)=0$. Note that $\mathcal{B}_{\mathrm{ww}, M}^{\min } \geq 1$.

- $\llbracket x\langle y\rangle \rrbracket_{\gamma}^{W}$ for S-box $x$ depends upon whether $x$ is injective, which is determined by checking whether some constant in the array appears more than once if the S-box is given by an array, or by checking the satisfiability of the constraint $x \neq x^{\prime} \wedge f(x)=f\left(x^{\prime}\right)$ (via SMT solving) if the S-box is defined by an $s_{-} f n$ function $f$. If it is injective, $\llbracket x\langle y\rangle \rrbracket_{\gamma}^{\mathrm{W}}$ gives the pair $\left(\left\{b=b^{\prime}\right\}, b^{\prime}\right)$, where the Boolean variable $b$ models the difference of $y$; the fresh Boolean variable $b^{\prime}$ models the difference of the result $x\langle y\rangle$; and the constraint $b=b^{\prime}$ ensures that $b=1$ iff $b^{\prime}=1$. If it is non-injective, the constraint $b \geq b^{\prime}$ is imposed instead of $b=b^{\prime}$, as $x\langle y\rangle$ may differ in two executions only if $y$ differs in the two executions.

Lemma 5.1. Suppose $\llbracket e \rrbracket_{\gamma}^{W}=(\Psi, \vec{b})$ with $\Psi \neq \emptyset . \vec{b}=\left(b_{1}, \cdots, b_{m}\right)$ is a solution of $\Psi$ if and only if $\left(b_{1}, \cdots, b_{i}\right)$ is feasible differences of the operands and result of e, where $\left(b_{i+1}, \cdots, b_{m}\right)$ is for the possible auxiliary Boolean variables.

### 5.2 Word-wise Differential Denotational Semantics for Statements

Denotational semantics for statements. The (word-wise differential) denotational semantics of a statement $S$ is given by $\llbracket S \rrbracket^{\mathrm{W}}$ that maps each state $\gamma \in \Gamma$ to a triple $\left(\Psi, \gamma^{\prime}, \Theta\right)$, denoted by $\llbracket S \rrbracket_{\gamma}^{\mathrm{W}}$, where $\Psi$ is defined as above (i.e., set of IL constraints), $\gamma^{\prime}$ is the updated state, and $\Theta$ is a set of Boolean variables each of which models the input difference of an $S$-box under two executions.
Denotational semantic rules. The semantics for statements in EASyBC is shown in Figure 7.
The semantic rules $\llbracket \tau x \rrbracket_{\gamma}^{\mathrm{W}}, \llbracket x[i]=y \rrbracket_{\gamma}^{\mathrm{W}}$ and $\llbracket x=e \rrbracket_{\gamma}^{\mathrm{W}}$ for declaration $\tau x$, array put $x[i]=y$, assignment $x=e$ are straightforward, which updates the state $\gamma$ accordingly to track the mapping from variables $x$ to Boolean variables $\gamma(x)$ that models the difference of $x$ under two executions, the set of constraints $\Psi$ is collected from the semantics $\llbracket e \rrbracket_{\gamma}^{W}$ of the expression $e$, and moreover, the Boolean variable $\gamma(y)$ modeling the difference of the input $y$ of an S-box is recorded in $\Theta$.

The semantic rule $\llbracket x=g\left(y_{1}, \cdots, y_{m}\right) \rrbracket_{\gamma}^{\mathrm{W}}$ for a function call $x=g\left(y_{1}, \cdots, y_{m}\right)$ follows its operational semantics. We first pass the Boolean variables $\gamma\left(y_{i}\right)$ for $1 \leq i \leq m$ that model the differences of the actual arguments $y_{i}$ to the formal parameters $x_{i}$, then iteratively evaluate each statement $S_{i}$ in its function body, and finally maps the variable $x$ to the Boolean variable $\gamma(y)$ that models the difference of the return $y$. The set $\Psi$ of IL constraints and the set $\Theta$ of Boolean variables modeling the input differences of S-boxes are collected from them of the statements, i.e., $\Psi_{i}$ 's and $\Theta_{i}$ 's

### 5.3 Word-wise Resistance Evaluation

To evaluate the resistance of the program $P$, we define the semantics $\llbracket P \rrbracket^{\mathrm{W}}$ of the program $P$ as the semantics of its function as follows:

$$
\llbracket P \rrbracket^{\mathbb{W}}=\llbracket \text { uints }[n] f\left(\text { uints }\left[n_{1}\right] k \text {, uints }[n] t x t\right)\left\{S_{1} ; \cdots ; S_{n} ; \text { return } y ;\right\} \rrbracket^{\mathbb{W}}=\left(\Psi, \gamma_{n}, \Theta\right)
$$

where $\Psi=\bigcup_{i=1}^{n} \Psi_{i}, \Theta=\bigcup_{i=1}^{n} \Theta_{i}, \llbracket S_{1} \rrbracket_{\gamma_{0}}^{W}=\left(\Psi_{1}, \gamma_{1}, \Theta_{1}\right), \cdots, \llbracket S_{n} \rrbracket_{\gamma_{n-1}}^{W}=\left(\Psi_{n}, \gamma_{n}, \Theta_{n}\right)$ and $\gamma_{0}$ is an initial state mapping each array element of the formal parameter $t x t$ to a fresh Boolean variable.

Clearly, $\Phi$ is the set of IL constraints imposed by the dependency of differences between support variables and results of all the operations in the program $P$, and $\sum_{b \in \Theta} b$ gives the sum of the number of active S-boxes under two executions of the program $P$. Determining the lower bound of the minimum number of active S-boxes under two executions of $P$ amounts to minimizing $\sum_{b \in \Theta} b$ subject to $\Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$, where $\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1$ ensures that the input difference of text $t x t$ is nonzero, otherwise $\sum_{b \in \Theta} b$ would trivially be 0 .

Recall from Section 3.5 that $\mathcal{N}_{\text {diff }}$ denotes the minimum number of active $S$-boxes in all the possible pairs of executions.

Theorem 5.2. Let $\llbracket P \rrbracket^{\mathrm{W}}=(\Phi, \gamma, \Theta)$ and $N$ be the minimum value of the objective function $\sum_{b \in \Theta} b$ subject to $\Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$. We have that $N \leq \mathcal{N}_{\text {diff }}$.

When the MILP program cannot be solved efficiently with large round number $s$, one can turn to the following decomposition approach.

Proposition 5.3. Let $n_{1}$ and $n_{2}$ be the minimum number of the active $S$-boxes of the first $r_{1}$-round and the subsequent $r_{2}$-round differential characteristics respectively, then $n_{1}+n_{2}$ is a lower bound of the minimum number of the active $S$-boxes of the $\left(r_{1}+r_{2}\right)$-round differential characteristics.

In practice, one may be only interested in proving the resistance against differential cryptanalysis instead of computing a bound. In this case, it suffices to prove that $\mathcal{N}_{\text {diff }} \geq \frac{-\hbar}{\log _{2} p}$, where $p$ denotes the maximum probability $\operatorname{Pr}_{\mathcal{S}}(\Delta X, \Delta Y)$ among all the nonzero differentials $(\Delta X, \Delta Y)$ for any Sbox $\mathcal{S}$ that is active in $s$-round differential characteristics and $\mathscr{G}$ is the block size of the cipher. By Theorem 5.2, we only need to verify if $\left\{\sum_{b \in \Theta} b<\frac{-\hbar}{\log _{2} p}\right\} \cup \Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$ is unsatisfiable.

Corollary 5.4. Let $\llbracket P \rrbracket^{W}=(\Phi, \gamma, \Theta)$. If $\left\{\sum_{b \in \Theta} b<\frac{-6}{\log _{2} p}\right\} \cup \Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$ is unsatisfiable, then the program $P$ with block size \& is resistant against differential cryptanalysis.

If the $s$-round cipher $P$ can be partitioned into $\frac{s}{s^{\prime}}$ identical $s^{\prime}$-round ciphers $P^{\prime}$, by Proposition 5.3 and Corollary 5.4, we can conclude that the cipher $P$ is resistant against differential cryptanalysis if the number of active S -boxes of the cipher $P^{\prime}$ is no less than $\frac{-s^{\prime} \cdot \epsilon}{s \cdot \log _{2} p}$.

Corollary 5.5. Let $\llbracket P^{\prime} \rrbracket^{\mathbb{W}}=(\Phi, \gamma, \Theta)$. If $\left\{\sum_{b \in \Theta} b<\frac{-s^{\prime} \cdot \mathscr{G}}{s \cdot \log _{2} p}\right\} \cup \Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$ is unsatisfiable, then the program $P$ with block size b is resistant against differential cryptanalysis.

## 6 BIT-WISE APPROACH

In this section, we present a bit-wise approach which, as the word-wise approach, determines the lower bound of the minimum number of active $S$-boxes, but lifts its limitation requiring that a program uses types uints[ $n$ ] for a fixed bit size $s$ of involved $S$-boxes and individual bits of any entry in an array cannot be changed.
High-level intuition. Fix a program $P$, which we normally assume cannot be handled by the wordwise approach. A straightforward idea is to transform the program $P$ to its Boolean counterpart $P^{\prime}$ by bit-blasting. However, this would introduce a large number of variables and statements, resulting in a prohibitively large MILP problem. In this work, we adopt a strategy to "implicitly" bit-blast, meaning that bit-blasting is performed in the generation of MILP. To this end, each bit of a variable in the program $P$ is modeled by one Boolean variable $b$, where $b=1$ in an MILP solution indicates that the corresponding bit differs in $P$ when executed under two distinct inputs for some fixed key. Thus, a variable of type uints[ $n]$ is modeled by a vector $\vec{b}$ of $s \cdot n$ Boolean variables. The word-wise differential denotational semantics is then lifted to the bit-wise one.

### 6.1 Bit-wise Differential Denotational Semantics for Expressions

State. We first lift the state $\gamma$ from word-wise to bit-wise. Let $\|x\|=s \cdot n$ for a variable $x$ of the type uints [ $n$ ]. A state $\gamma$ now maps each pair $(x, i) \in\left(\mathbb{X}_{f} \backslash \mathbb{K}_{f}\right) \times \mathbb{N}$ to a Boolean variable, where

- for each variable $x$ of type uints $[n], \gamma(x, i \cdot s+j)$ gives a Boolean variable modeling the difference of the $(j+1)$-th most significant bit of the $(i+1)$-th entry $x_{i}$ in the array $x$;
- $\gamma(x)$ denotes the sequence $\gamma(x, 0), \gamma(x, 1), \cdots, \gamma(x,\|x\|-1), \gamma[(x, i) \mapsto b]$ denotes the update of the state $\gamma$ by mapping $(x, i)$ to the Boolean variable $b$, and $\gamma[x \mapsto \vec{b}]$ denotes the update $\gamma\left[(x, 0) \mapsto \vec{b}_{0}\right] \cdots\left[(x,\|x\|-1) \mapsto \vec{b}_{\|x\|-1}\right]$ for a Boolean vector $\vec{b}$ with $\|x\|=\|\vec{b}\|$.
Denotational semantics. The (bit-wise differential) denotational semantics of an expression $e$ is given by $\llbracket e \rrbracket^{\mathrm{B}}$ that maps each state $\gamma \in \Gamma$ to a pair $(\Psi, \vec{b})$, denoted by $\llbracket e \rrbracket_{\gamma}^{\mathrm{B}}$, where
- $\Psi$ is a set of IL constraints over Boolean variables characterizing the bit-level dependency of differences between the support variables and result of $e$ such that the differences are a solution of $\Psi$ iff these bit-level differences are feasible for support variables and result of $e$;
- $\vec{b}$ such that $\|\vec{b}\|=\|e\|$ is a Boolean vector modeling the bit-level differences of the result of $e$.

Denotational semantic rules. The semantics for expressions in EAsYBC is shown in Figure 8. The semantic rules $\llbracket \sim x \rrbracket_{\gamma}^{\mathrm{B}}, \llbracket \operatorname{View}(x, i, j) \rrbracket_{\gamma}^{\mathrm{B}}, \llbracket \operatorname{touint}\left(x_{1}, \cdots, x_{m}\right) \rrbracket_{\gamma}^{\mathrm{B}}$ and $\llbracket x\langle\cdot y \cdot\rangle \rrbracket_{\gamma}^{\mathrm{W}}$ are trivial according to their operational semantics. Below, we explain the other non-trivial ones.

- The semantic rule $\llbracket x \odot y \|_{\gamma}^{B}$ for $\odot \in\{\wedge, \vee\}$ gives the pair $\left(\left\{\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \vec{b}_{i}^{0} \mid 0 \leq i<\|x\|\right\}, \vec{b}^{0}\right)$, where for every $0 \leq i<\|x\|, \vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \vec{b}_{i}^{0}$ characterizes that if the $(i+1)$-th bits of the operands $x$ and $y$ have no differences (i.e., $\vec{b}_{i}^{1}=\vec{b}_{i}^{2}=0$ ), then the $(i+1)$-th bit of the result $x \odot y$ has no differences (i.e., $\vec{b}_{i}^{0}=0$ ). Otherwise, it may have differences (with the probability of $\frac{1}{2}$ ).
- The semantic rule $\llbracket x \oplus y \rrbracket_{\gamma}^{\mathrm{B}}$ gives the pair $\left(\bigcup_{i=0}^{\|x\|-1} \psi_{\oplus}^{j}\left(\vec{b}_{i}^{1}, \vec{b}_{i}^{2}, \vec{b}_{i}^{0}\right), \vec{b}^{0}\right)$, where for each $0 \leq i<\|x\|$, $\psi_{\oplus}^{j}\left(\vec{b}_{i}^{1}, \vec{b}_{i}^{2}, \vec{b}_{i}^{0}\right)$ characterizes that either the $(i+1)$-th bits of the operands $x, y$ and result $x \oplus y$ have no differences (i.e., $\vec{b}_{i}^{0}=\vec{b}_{i}^{1}=\vec{b}_{i}^{2}=0$ ) or exactly two of them have differences (i.e., $\vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2}=2$ ).
- The semantic rule $\llbracket x \odot y \rrbracket_{\gamma}^{\mathrm{B}}$ for $\odot \in\{+,-\}$ gives the pair $\left(\bigcup_{i=0}^{\|x\|-1} \Psi_{i}, \vec{b}^{0}\right)$, where $\Psi_{i}$ characterizes the dependency of the differences between the $i$-th and $(i+1)$-th bits of the operands $x, y$ and result $x \odot y$. The dependency is obtained by bit-blasting $x+y$ via a ripple-carry adder, i.e.,
$-\operatorname{bin}_{i}(x+y)=\operatorname{bin}_{i}(x) \oplus \operatorname{bin}_{i}(y) \oplus \vec{c}_{i}$, for every $0 \leq i<\|x\|$;
- the carry bit $\vec{c}_{i}=1$ iff $\operatorname{bin}_{i-1}(x)+\operatorname{bin}_{i-1}(y)+\vec{c}_{i-1} \geq 2$, for every $1 \leq i<\|x\|$, with $\vec{c}_{0}=0$, where $\operatorname{bin}_{i}(x)$ denotes the $(i+1)$-th most significant bit of $x$. Clearly, the difference $\vec{b}_{i}^{0}$ of the bit $\operatorname{bin}_{i}(x+y)$ depends upon the differences $\vec{b}_{i-1}^{1}, \vec{b}_{i-1}^{2}, \vec{b}_{i}^{1}$ and $\vec{b}_{i}^{2}$ of the bits $\operatorname{bin}_{i-1}(x), \operatorname{bin}_{i-1}(y)$, $\operatorname{bin}_{i}(x)$ and $\operatorname{bin}_{i}(y)$. Indeed, we can deduce the dependency:
- if $\vec{b}_{i-1}^{0}=\vec{b}_{i-1}^{1}=\vec{b}_{i-1}^{2}=1$, then $\vec{c}_{i-1} \oplus \vec{c}_{i-1}^{\prime}=\vec{b}_{i-1}^{3}=\vec{b}_{i}^{3}=1$ and $\vec{b}_{i}^{0}=\neg\left(\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}\right)$,
- if $\vec{b}_{i-1}^{0}=\vec{b}_{i-1}^{1}=\vec{b}_{i-1}^{2}=0$, then $\vec{c}_{i-1} \oplus \vec{c}_{i-1}^{\prime}=\vec{b}_{i-1}^{3}=\vec{b}_{i}^{3}=0$ and $\vec{b}_{i}^{0}=\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}$,
- otherwise $1 \leq \vec{b}_{i-1}^{0}+\vec{b}_{i-1}^{1}+\vec{b}_{i-1}^{2} \leq 2$. Indeed, the probability of $\vec{b}_{i}^{0}=\neg\left(\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}\right)$, (resp. $\vec{b}_{i}^{0}=\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}$ and $\vec{b}_{i}^{0}=1$ ) is $\frac{1}{2}$.
The above dependency is characterized by the set of IL constraints $\Psi_{i}$. Furthermore, $\Psi_{0}$ and $\Psi_{1}$ can be simplified by $\vec{c}_{0}=0$. The semantic rule $\llbracket x-y \rrbracket_{\gamma}^{\mathrm{B}}$ is defined the same as $\llbracket x+y \rrbracket_{\gamma}^{\mathrm{B}}$ because $z=x-y$ iff $x=z+y$, and the Boolean variables $\vec{b}_{i}^{0}, \vec{b}_{i}^{1}$ and $\vec{b}_{i}^{2}$ in $\Psi_{i}$ are symmetric.
- The semantic rule $\llbracket M * x \rrbracket_{\gamma}^{\mathrm{B}}$ gives the pair $\left(\bigcup_{i=0}^{|x|-1} \bigcup_{h=0}^{s-1} \Psi_{M, \mathrm{i}, \mathrm{h}}^{v}, \overrightarrow{b^{\prime}}\right)$, where for each $0 \leq i<|x|$ and each $0 \leq h<s, \Psi_{M, i, h}^{v}$ characterizes the bit-level dependency of the differences between the

|  |  |
| :---: | :---: |
| $\llbracket \operatorname{View}(x, i, j) \rrbracket_{\gamma}^{\mathrm{B}}=\left(\emptyset,\left(\vec{b}_{i \cdot s+0}, \cdots, \vec{b}_{i \cdot s+s-1}, \cdots, \vec{b}_{j \cdot s+0}, \cdots, \vec{b}_{j \cdot s+s-1}\right)\right)$, where $\vec{b}=\gamma(x)$ |  |
| $\llbracket x \wedge y \rrbracket_{\gamma}^{\mathrm{B}}=\llbracket x \vee y \rrbracket_{\gamma}^{\mathrm{B}}=\left(\left\{\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \vec{b}_{i}^{0} \mid 0 \leq i<\\|x\\|\right\}, \vec{b}^{0}\right)$, where $\vec{b}^{1}=\gamma(x), \vec{b}^{2}=\gamma(y), \vec{b}^{0}=$ newBV() |  |
| $\begin{aligned} & \llbracket x \oplus y \\|_{Y}^{B}=\left(\cup_{i=0}^{\\|x\\|-1} \psi_{\oplus}^{j}\left(\vec{b}_{i}^{1}, \vec{b}_{i}^{2}, \vec{b}_{i}^{0}\right), \vec{b}^{0}\right), \text { where } \vec{b}^{1}=\gamma(x), \vec{b}^{2}=\gamma(y), \vec{b}^{0}=\text { newBV }(), b^{\prime}=\text { newBV }() \\ & \psi_{\oplus}^{1}\left(b^{1}, b^{2}, b^{0}\right)=\left\{2 \geq b^{0}+b^{1}+b^{2} \geq 2 b^{\prime}, b^{\prime} \geq b^{0}, b^{\prime} \geq b^{1}, b^{\prime} \geq b^{2}\right\}[\text { Sun et al. 2014a] } \\ & \psi_{\oplus}^{2}\left(b^{1}, b^{2}, b^{0}\right)=\left\{b^{0}+b^{1}+b^{2} \leq 2, b^{1}+b^{2} \geq b^{0}, b^{0}+b^{1} \geq b^{2}, b^{0}+b^{2} \geq b^{1}\right\} \text { [Sasaki and Todo 2017] } \\ & \psi_{\oplus}^{3}\left(b^{1}, b^{2}, b^{0}\right)=\left\{b^{0}+b^{1}+b^{2}=2 b^{\prime}\right\} \text { [Cui et al. 2016] } \end{aligned}$ |  |
| $\begin{aligned} & \llbracket x+y \rrbracket_{\gamma}^{\mathrm{B}}=\llbracket x-y \rrbracket_{\gamma}^{\mathrm{B}}=\left(\bigcup_{i=0}^{\\|x\\|-1} \Psi_{i}, \vec{b}^{0}\right), \text { where } \vec{b}^{1}=\gamma(x), \vec{b}^{2}=\gamma(y), \vec{b}^{0}=\text { newBV }(), \\ & \Psi_{0}=\Psi_{\oplus}^{i}\left(\vec{b}_{0}^{0}, \vec{b}_{0}^{1}, \vec{b}_{0}^{2}\right) \quad \Psi_{1}=\left\{\begin{array}{r} \vec{b}_{1}^{0}+\vec{b}_{1}^{1}+\vec{b}_{1}^{2} \leq \vec{b}_{0}^{1}+\vec{b}_{0}^{2}+2, \quad-\vec{b}_{1}^{0}+\vec{b}_{1}^{1}-\vec{b}_{1}^{2} \leq \vec{b}_{0}^{1}+\vec{b}_{0}^{2}, \\ -\vec{b}_{1}^{0}-\vec{b}_{1}^{1}+\vec{b}_{1}^{2} \leq \vec{b}_{0}^{1}+\vec{b}_{0}^{2}, \quad \vec{b}_{1}^{0}-\vec{b}_{1}^{1}-\vec{b}_{1}^{2} \leq \vec{b}_{0}^{1}+\vec{b}_{0}^{2} \end{array}\right\} \\ & \forall 2 \leq i<\\|x\\| . \Psi_{i}=\left\{\begin{array}{ll} 4 \geq \sum_{j=0}^{2} \vec{b}_{i-1}^{j}-\vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq 0, & 4 \geq \sum_{j=0}^{2} \vec{b}_{b i-1}^{j}+\vec{b}_{i}^{0}+\vec{b}_{i}^{1}-\vec{b}_{i}^{2} \geq 0, \\ 4 \geq \sum_{j=0}^{2} \vec{b}_{i-1}^{j}+\vec{b}_{i}^{0}-\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq 0, & \sum_{j=0}^{2} \vec{b}_{i-1}^{j}+2 \geq \vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \sum_{j=0}^{2} \vec{b}_{i-1}^{j}-2 \end{array}\right\} \end{aligned}$ |  |
| $\llbracket M * x \rrbracket_{\gamma}^{B}=\left(\cup_{i=0}^{\|x\|-1} \cup_{h=0}^{s-1} \Psi_{M, i, h}^{v}, \overrightarrow{b^{\prime}}\right)$, where $\vec{b}=\gamma(x), \vec{b}^{\prime}=\operatorname{newBV}(), b_{\text {new }}^{i}=\operatorname{newBV}(), v \in\{1,2\}$ $\Psi_{M, i, \mathrm{~h}}^{v}=\psi_{\oplus}^{v}\left(b^{0}, b^{1}, b_{\text {new }}^{1}\right) \cup \psi_{\oplus}^{v}\left(b_{\text {new }}^{1}, b^{2}, b_{\text {new }}^{2}\right) \cup \cdots \cup \psi_{\oplus}^{v}\left(b_{\text {new }}^{m-2}, b^{m-1}, b_{\text {new }}^{m-1}\right) \cup \psi_{\oplus}^{v}\left(b_{\text {new }}^{m-1}, b^{m}, \vec{b}_{i \cdot s+h}^{\prime}\right)$ <br> $\Psi_{\mathrm{M}, \mathrm{i} \mathrm{h}}^{3}=\left\{\vec{b}_{i \cdot \mathrm{~s}+h}^{\prime}+\sum_{j=0}^{m} b^{j}=2 d, 0 \leq d \leq\left\lfloor\frac{m+2}{2}\right\rfloor\right\}$, where $d$ is a fresh integer variable, $\left\{b^{0}, \cdots, b^{m}\right\}$ is the set of support variables of $\left(\bigoplus_{0 \leq j<\|x\|, h \leq k<s, M_{i, j, k}=1} \vec{b}_{j \cdot s+k}\right) \oplus\left(\vec{c}_{h} \wedge \bigoplus_{0 \leq j<\|x\|, 0 \leq k<\left\lfloor\frac{s}{2}\right\rfloor} \vec{b}_{j \cdot s+2 k}\right)$ |  |
| $\llbracket x\left(\cdot y \cdot \\|_{\gamma}^{\mathrm{B}}=\left(\emptyset, \vec{b}^{0}\\|\cdots\\|\|\vec{b}\| x \mid-1\right)\right.$, where $\vec{b}=\gamma(y), x$ is P-box $\left(j_{0}, \cdots, j_{n-1}\right)$, and $\vec{b}^{i}=\left(\vec{b}_{j_{i} \cdot s+0}, \cdots, \vec{b}_{j_{i} \cdot s+s-1}\right)$ for $0 \leq i \leq\|x\|-1$ |  |
| $\begin{array}{r} \\|x\langle y\rangle\\|_{\gamma}^{\mathbb{D}}=\left(\Psi_{\mathcal{S}} \cup\right. \\ \text { and } \end{array}$ | $\begin{aligned} & x: \mathbb{B}^{n} \rightarrow \mathbb{B}^{m} \text { is an S-box } \mathcal{S}, \vec{b}^{\prime}=\text { newBV }(), b=\text { newBV } \\ & \vec{b}_{i}+\sum_{i=0}^{m-1} \vec{b}_{i}^{\prime} \leq \mathcal{B}_{\mathrm{bw}, x^{\prime}}^{\max }, b \geq \vec{b}_{i}, b \geq \vec{b}_{j}^{\prime} \mid 0 \leq i<n, 0 \leq j \end{aligned}$ |

Fig. 8. The bit-wise differential denotational semantic rules for expressions, where $s=\frac{\|x\|}{|x|}$ denotes the bit width of the entries of the array $x$, and $\vec{c}=\operatorname{bin}\left(2 \otimes 2^{s-1}\right)$ is the $s$-bitstream corresponding to the coefficients of the irreducible polynomial for the underlying finite-field.
array $x$ and the $(h+1)$-th $\operatorname{bit}^{\operatorname{bin}}{ }_{h}\left(y_{i}\right)$ of the $(i+1)$-th entry $y_{i}$ in the resulting array $y=M * x$. The dependency is obtained by expanding the matrix-vector product as follows:

$$
M * x=\left(\bigoplus_{j=0}^{|x|-1}\left(M_{0, j} \otimes x_{j}\right), \cdots, \bigoplus_{j=0}^{|x|-1}\left(M_{|x|-1, j} \otimes x_{j}\right)\right) .
$$

Clearly, the $(i+1)$-th entry $y_{i}$ is $\bigoplus_{j=0}^{|x|-1}\left(M_{i, j} \otimes x_{j}\right)$ and thus the difference $\vec{b}_{i \cdot s+h}^{\prime}$ of its $(h+1)$-th bit $\operatorname{bin}_{h}\left(y_{i}\right)$ is the parity of the differences of the $(h+1)$-th bits in $M_{i, j} \otimes x_{j}$ for $0 \leq j<h$. The expression $M_{i, j} \otimes x_{j}$ is bit-blasted by expanding the finite-field multiplication ( $\otimes$ ) using a series of modular left shifts, XOR operations and the coefficients $\vec{c}$ of the irreducible polynomial for the underlying finite field, where $\vec{c}=\operatorname{bin}\left(2 \otimes 2^{s-1}\right)$. We finally can deduce that the parity of the differences of the $(h+1)$-th bits in $M_{i, j} \otimes x_{j}$ for $0 \leq j<h$ is

$$
\left(\bigoplus_{0 \leq j<|x|, h \leq k<s, M_{i, j, k}=1} \vec{b}_{j \cdot s+k}\right) \oplus\left(\vec{c}_{h} \wedge \bigoplus_{0 \leq j<|x|, 0 \leq k<\left\lfloor\frac{s}{2}\right\rfloor} \vec{b}_{j \cdot s+2 k}\right) .
$$

Let $\left\{b^{0}, \cdots, b^{m}\right\}$ be the set of support variables of the above expression. We have:

$$
\vec{b}_{i \cdot s+h}^{\prime}=\bigoplus_{t=0}^{m} b_{t},
$$

which can be alternatively characterized by $\Psi_{M, \mathrm{i}, \mathrm{h}}^{v}$ for $v \in\{1,2,3\}$.

- The semantic rule $\llbracket x\langle y\rangle \rrbracket_{\gamma}^{W}$ for S-box $x=\mathcal{S}$ gives the pair $\left(\Psi_{\mathcal{S}} \cup \Psi_{\mathcal{S}}^{\mathrm{bn}}, \vec{b}^{\prime}\right)$, where $\Psi_{\mathcal{S}}$ is a set of IL constraints characterizing the bit-level dependency of differences between the input $y$ and the result $\mathcal{S}(y)$ (cf. Section 4.2) and $\Psi_{S}^{\text {bn }}$ enforces that the Hamming weight of the bitstream $\vec{b} \| \vec{b}^{\prime}$ ranges from $\mathcal{B}_{\mathrm{bw}, \mathcal{S}}^{\min }$ to $\mathcal{B}_{\mathrm{bw}, \mathcal{S}}^{\max }$. Though $\Psi_{\mathcal{S}}^{\mathrm{bn}}$ is redundant, it often boosts MILP solving.

Lemma 6.1. Suppose $\llbracket e \rrbracket_{\gamma}^{\mathrm{B}}=(\Psi, \vec{b})$ with $\Psi \neq \emptyset$. The assignment $\left(b_{1}, \cdots, b_{m}\right)$ is a solution of $\Psi$ if and only if $\left(b_{1}, \cdots, b_{i}\right)$ is feasible bit-level differences of the operands and result of e, where $\left(b_{i+1}, \cdots, b_{m}\right)$ is for the possible auxiliary Boolean variables.

### 6.2 Bit-wise Differential Denotational Semantics for Statements

The (bit-wise differential) denotational semantics of a statement $S$ is given by $\llbracket S \|^{\mathrm{B}}$ that maps each state $\gamma \in \Gamma$ to a triple $\left(\Psi, \gamma^{\prime}, \Theta\right)$, denoted by $\llbracket S \rrbracket_{\gamma}^{B}$, where $\Psi, \gamma^{\prime}$ and $\Theta$ are the same as above.

The bit-wise differential semantic rules for statements in EASYBC are the same as those word-wise ones except for array put and S-box access, which are given below:

| $\vec{b}=\gamma(y) \quad x$ has type uints $[n] \quad \gamma^{\prime}=\gamma\left[(x, i \cdot s+0) \mapsto \vec{b}_{0}\right] \cdots\left[(x, i \cdot s+s-1) \mapsto \vec{b}_{s-1}\right]$ |
| :---: |
| $\llbracket x[i]=y \rrbracket_{\gamma}^{\mathrm{B}}=\left(\emptyset, \gamma^{\prime}, \emptyset\right)$ |
| $\llbracket x^{\prime}\langle y\rangle \\|_{\gamma}^{\mathrm{B}}=(\Psi, \vec{b}) \quad \vec{b}^{\prime}=\gamma(y) \quad b_{x}=\operatorname{newBV}()$ |
| $\Psi^{\prime}=\Psi \cup\left\{\sum_{j=0}^{\\|y\\|-1} \vec{b}_{j}^{\prime} \geq b_{x} \geq \vec{b}_{i}^{\prime} \mid 0 \leq i<\\|y\\| \quad \gamma^{\prime}=\gamma[x \mapsto \vec{b}]\right\}$ |
| $\left\lfloor x=x^{\prime}\langle y\rangle \rrbracket_{\gamma}^{\mathrm{B}}=\left(\Psi^{\prime}, \gamma^{\prime},\left\{b_{x}\right\}\right)\right.$ |

Intuitively, the semantic rule $\llbracket x[i]=y \rrbracket_{\gamma}^{\mathrm{B}}$ updates the state $\gamma$ accordingly by mapping the bits of the $(i+1)$-th entry in the array $x$ to the Boolean variables $\vec{b}$ that model the differences of the bits of the result $e$. The semantic rule $\llbracket x=x^{\prime}\langle y\rangle \rrbracket_{\gamma}^{\mathrm{B}}$ adds a fresh Boolean variable $b_{x}$, where if $b_{x}=1$, then the S-box is active, i.e., some bit $\vec{b}_{i}^{\prime}$ that models the difference of one bit of input $y$ is nonzero.

### 6.3 Bit-wise Resistance Evaluation

To evaluate the resistance of the program $P$ in a bit-wise manner, similar to $\llbracket P \rrbracket^{W}$ (cf. Section 5.3), we define the (bit-wise) semantics $\llbracket P \rrbracket^{B}$ of the program $P$ using its fn function $f$ as follows.

$$
\llbracket P \rrbracket^{\mathrm{B}}=\llbracket \text { uints }[n] f\left(\text { uints }\left[n_{1}\right] k, \text { uints }[n] t x t\right)\left\{S_{1} ; \cdots ; S_{n} ; \text { return } y ;\right\} \rrbracket^{\mathrm{B}}=\left(\Psi, \gamma_{n}, \Theta\right)
$$

where $\Psi=\bigcup_{i=1}^{n} \Psi_{i}, \Theta=\bigcup_{i=1}^{n} \Theta_{i}, \llbracket S_{1} \rrbracket_{\gamma_{0}}^{\mathrm{B}}=\left(\Psi_{1}, \gamma_{1}, \Theta_{1}\right), \cdots, \llbracket S_{n} \rrbracket_{\gamma_{n-1}}^{\mathrm{B}}=\left(\Psi_{n}, \gamma_{n}, \Theta_{n}\right)$ and $\gamma_{0}$ is an initial state mapping each bit of array elements of $t x t$ to a fresh Boolean variable. We get that:

Theorem 6.2. Let $\llbracket P \rrbracket^{\mathrm{B}}=(\Phi, \gamma, \Theta)$ and $N$ be the minimum value of the objective function $\sum_{b \in \Theta} b$ subject to the set of IL constraints $\Phi \cup\left\{\sum_{i=0}^{s \cdot n-1} \gamma(t x t, i) \geq 1\right\}$. We have that $N \leq \mathcal{N}_{\text {diff }}$.

Corollary 5.4 and Corollary 5.5 still hold when $\llbracket P \rrbracket^{\mathrm{W}}=(\Phi, \gamma, \Theta)$ is replaced by $\llbracket P \rrbracket^{\mathrm{B}}=(\Phi, \gamma, \Theta)$.

## 7 EXTENDED BIT-WISE APPROACH

The lower bound of the minimum number of the active S-boxes is often effective, but it may not be sufficiently tight to prove the resistance [Sun et al. 2014b], as the probabilities between input and output differences for the operations $\wedge, \vee,+,-$ and S-boxes are not fully addressed. Furthermore, the bit-wise approach is not applicable if non-linear layers are implemented in other operations than S-boxes (e.g., SIMON), or S-boxes are too large to be given as arrays (e.g., SPARKLE). In this section, we extend the bit-wise approach to directly bound the MaxEDCP rather than by bounding the minimum number of the active $S$-boxes.

$$
\begin{aligned}
& \llbracket x \wedge y \rrbracket_{\gamma}^{\mathrm{EB}}=\llbracket x \vee y \rrbracket_{\gamma}^{\mathrm{EB}}=\left(\bigcup_{i=0}^{\|x\|-1} \Psi_{i}, \vec{b}^{0}, \sum_{i=0}^{\|x\|-1} \vec{p}_{i}\right) \text {, where } \vec{b}^{1}=\Gamma(x), \vec{b}^{2}=\Gamma(y), \vec{b}^{0}=\text { newBV }() \\
& \Psi_{i}=\left\{\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \vec{p}_{i}, \vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \leq 3 \vec{p}_{i}\right\} \\
& \llbracket x+y \rrbracket_{\gamma}^{\mathrm{EB}}=\llbracket x-y \rrbracket_{\gamma}^{\mathbb{E B}}=\left(\cup_{i=0}^{\|x\|-1} \Psi_{i}, \vec{b}^{0}, \sum_{i=1}^{\|x\|-1} \vec{p}_{i}\right) \text {, where } \vec{b}^{1}=\Gamma(x), \vec{b}^{2}=\Gamma(y), \vec{b}^{0}=\text { newBV }() \\
& \Psi_{0}=\Psi_{\oplus}^{i}\left(\vec{b}_{0}^{0}, \vec{b}_{0}^{1}, \vec{b}_{0}^{2}\right) \quad \Psi_{1}=\left\{\begin{array}{r}
\vec{b}_{1}^{0}+\vec{b}_{1}^{1}+\vec{b}_{1}^{2} \leq \vec{p}_{1}+2,-\vec{b}_{1}^{0}+\vec{b}_{1}^{1}-\vec{b}_{1}^{2} \leq \vec{p}_{1},-\vec{b}_{1}^{0}-\vec{b}_{1}^{1}+\vec{b}_{1}^{2} \leq \vec{p}_{1}, \\
\vec{b}_{1}^{0}-\vec{b}_{1}^{1}-\vec{b}_{1}^{2} \leq \vec{p}_{1}, \vec{p}_{1} \leq \vec{b}_{0}^{1}+\vec{b}_{0}^{2} \leq 2 \vec{p}_{1}
\end{array}\right\} \\
& \forall 2 \leq i<\|x\| . \Psi_{i}=\Psi_{i}^{1} \cup \Psi_{i}^{2} \quad \Psi_{i}^{1}=\left\{\begin{array}{r}
3-\vec{p}_{i} \geq \sum_{j=0}^{2} \vec{b}_{i-1}^{j} \geq \vec{p}_{i}, \vec{p}_{i} \geq \vec{b}_{i-1}^{0}-\vec{b}_{i-1}^{1} \\
\vec{p}_{i} \geq \vec{b}_{i-1}^{1}-\vec{b}_{i-1}^{2}, \quad \vec{p}_{i} \geq \vec{b}_{i-1}^{2}-\vec{b}_{i-1}^{0}
\end{array}\right\} \\
& \Psi_{i}^{2}=\left\{\begin{array}{ll}
2-\vec{b}_{i-1}^{1}+\vec{p}_{i} \geq-\vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq-\vec{b}_{i-1}^{1}-\vec{p}_{i}, & 2-\vec{b}_{i-1}^{1}+\vec{p}_{i} \geq \vec{b}_{i}^{0}+\vec{b}_{i}^{1}-\vec{b}_{i}^{2} \geq-\vec{b}_{b}^{1}-\vec{p}_{i}, \\
2-\vec{b}_{i-1}^{1}+\vec{p}_{i} \geq \vec{b}_{i}^{0}-\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq-\vec{b}_{i-1}^{1}-\vec{p}_{i}, & 2+\vec{b}_{i-1}^{1}+\vec{p}_{i} \geq \vec{b}_{i}^{0}+\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq \vec{b}_{i-1}^{1}-\vec{p}_{i}
\end{array}\right\} \\
& \llbracket x\langle y\rangle \rrbracket_{Y}^{\mathrm{EB}}=\left(\Psi_{\mathcal{S}}^{\dagger} \cup \Psi_{\mathcal{S}}^{\mathrm{bn}}, \vec{b}^{\prime}, \sum_{j=1}^{k} t_{i_{j}} \cdot p_{i_{j}}\right) \text {, where } x \text { is an S-box } \mathcal{S}: \mathbb{B}^{n} \rightarrow \mathbb{B}^{m}
\end{aligned}
$$

Fig. 9. The extended bit-wise differential denotational semantic rules for expressions, where $\vec{p}=$ new $B V()$ is a vector of fresh Boolean variables used to encode probabilities, $\Psi_{\oplus}^{i}\left(\vec{b}_{0}^{0}, \vec{b}_{0}^{1}, \vec{b}_{0}^{2}\right)$ for $i \in\{1,2,3\}$ and $\Psi_{\mathcal{S}}^{\text {bn }}$ are defined in Figure 8, $\Psi_{\mathcal{S}}^{\dagger}$ and $\sum_{j=1}^{k} t_{i_{j}} \cdot p_{i_{j}}$ are defined in Section 4.3.

High-level intuition. Apart from the encoding presented in Section 4.3 for S-boxes, we further encode the probabilities between input and output differences for the operations $\wedge, ~ \vee,+,-$ into IL constraints using additional Boolean variables. The weighted sum $\varrho$ of these additional Boolean variables retains the probability $2^{-\varrho}$. Thus, the objective function is designed to minimize $\varrho$ instead of the number of the active $S$-boxes.

### 7.1 Extended Bit-wise Differential Denotational Semantics for Expressions

Denotational semantics. The (extended bit-wise differential) denotational semantics of an expression $e$ is given by $\llbracket e \rrbracket^{E B}$ that maps each state $\gamma \in \Gamma$ to a triple $(\Psi, \vec{b}, \varrho)$, denoted by $\llbracket e \rrbracket_{\gamma}^{E B}$, where the state $\gamma$, set of IL constraint $\Psi$ and Boolean vector $\vec{b}$ are the same as in Section 6.1, and the expression $\varrho$ is a weighted sum of Boolean variables encoding the probability $2^{-\varrho}$.
Denotational semantic rules. The semantic rules $\llbracket e \rrbracket_{\gamma}^{E B}$ for $\wedge, \vee,+,-$ and S-boxes are shown in Figure 9 , otherwise $\llbracket e \rrbracket_{\gamma}^{\mathrm{EB}}=(\Psi, \vec{b}, 0)$ if $\llbracket e \rrbracket_{\gamma}^{\mathrm{B}}=(\Psi, \vec{b})$, meaning that the probability between the input and output differences characterized by $\Psi$ is 1 (i.e., $2^{-0}$ ).

- The semantic rule $\llbracket x \odot y \rrbracket_{\gamma}^{\text {EB }}$ for $\odot \in\{\wedge, \vee\}$ gives the triple $\left(\bigcup_{i=0}^{\|x\|-1} \Psi_{i}, \vec{b}^{0}, \sum_{i=0}^{\|x\|-1} \vec{p}_{i}\right)$, where for every $0 \leq i<\|x\|, \Psi_{i}$ ensures that the probability of the difference $\vec{b}_{i}^{0}$ of the $(i+1)$-th bit in the result $x \odot y$ is $2^{-\vec{p}_{i}}$ when the differences of the $(i+1)$-th bits of the operands $x$ and $y$ are $\vec{b}_{i}^{1}$ and $\vec{b}_{i}^{2}$, respectively. Indeed, the probability of $\vec{b}_{i}^{0}=0$ is $2^{-0}$ when $\vec{b}_{i}^{1}=\vec{b}_{i}^{2}=0$, then $\vec{p}_{i}$ must be 0 . The probability of $\vec{b}_{i}^{0}=1$ is $2^{-1}$ when $\vec{b}_{i}^{1}+\vec{b}_{i}^{2} \geq 1$, then $\vec{p}_{i}$ must be 1 .
- The semantic rule $\llbracket x \odot y \rrbracket_{\gamma}^{\text {EB }}$ for $\odot \in\{+,-\}$ gives the triple $\left(\bigcup_{i=0}^{\|x\|-1} \Psi_{i}, \vec{b}^{0}, \sum_{i=1}^{\|x\|-1} \vec{p}_{i}\right)$, where for every $0 \leq i<\|x\|, \Psi_{i}$ ensures that for any fixed $\vec{b}_{i}^{1}$ and $\vec{b}_{i}^{2}$,
- if $\vec{b}_{i-1}^{0}=\vec{b}_{i-1}^{1}=\vec{b}_{i-1}^{2}, \vec{p}_{i}=0$ (i.e., the probability $2^{-\vec{p}_{i}}$ of $\vec{b}_{i}^{0}=\neg\left(\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}\right)$ or $\vec{b}_{i}^{0}=\vec{b}_{i}^{1} \oplus \vec{b}_{i}^{2}$ is 1 ),
- if $1 \leq \sum_{j=0}^{2} \vec{b}_{i-1}^{j} \leq 2, \vec{p}_{i}=1$ (i.e., the probability $2^{-\vec{p}_{i}}$ of $\vec{b}_{i}^{0}=1$ or $\vec{b}_{i}^{0}=0$ is $\frac{1}{2}$ ).

We remark that $\Psi_{0}$ and $\Psi_{1}$ can be simplified by $\vec{c}_{0}=0$, and the semantic rule $\llbracket x-y \rrbracket_{\gamma}^{\mathbb{E B}}$ is defined the same as $\llbracket x+y \rrbracket_{\gamma}^{\text {EB }}$ because $z=x-y$ iff $x=z+y$.

- The semantic rule $\llbracket x\langle y\rangle \rrbracket_{\gamma}^{\mathrm{EB}}$ follows the result given in Section 4.3, namely, $\left(\vec{b}, \vec{b}^{\prime}, p_{i_{1}}, \cdots, p_{i_{k}}\right)$ is a solution of $\Psi_{\mathcal{S}}^{\dagger}$ iff the probability $\operatorname{Pr}_{\mathcal{S}}\left(\vec{b}, \vec{b}^{\prime}\right)$ is $2^{-\sum_{j=1}^{k} t_{i j} \cdot p_{i_{j}}}$. Note that $t_{i_{j}}$ 's are constants and $\Psi_{\mathcal{S}}^{\mathrm{bn}}$ is added to boost MILP solving.

Lemma 7.1. Suppose $\llbracket e \rrbracket_{\gamma}^{\mathrm{EB}}=(\Psi, \vec{b}, \varrho)$ with $\Psi \neq \emptyset$. The assignment $\left\{b_{1}, \cdots, b_{m}, p_{1}, \cdots, p_{n}\right\}$ is a solution of $\Psi$ iff the probability of $\left\{b_{1}, \cdots, b_{i}\right\}$ being bit-level differences of the operands and result of e is $2^{-\varrho\left[p_{1}, \cdots, p_{n}\right]}$, where $\varrho\left[p_{1}, \cdots, p_{n}\right]$ denotes the value of $\varrho$ under the assignment $\left\{p_{1}, \cdots, p_{n}\right\}$ of the Boolean variables for encoding probabilities, and $\left\{b_{i+1}, \cdots, b_{m}\right\}$ is the assignment of the auxiliary Boolean variables if exist.

### 7.2 Extended Bit-wise Differential Denotational Semantics for Statements

The (extended bit-wise differential) denotational semantics of a statement $S$ is given by $\llbracket S \rrbracket^{\text {EB }}$ that maps each state $\gamma \in \Gamma$ to a triple ( $\Psi, \gamma^{\prime}, \varrho$ ), denoted by $\llbracket S \rrbracket_{\gamma}^{\mathrm{EB}}$, where $\Psi, \gamma^{\prime}$ and $\varrho$ are the same as above. The extended bit-wise differential semantic rules for statements in EasyBC are similar to those word-wise ones, where $\llbracket S \rrbracket_{\gamma}^{\mathrm{EB}}=\left(\Psi, \gamma^{\prime}, 0\right)$ if $\llbracket S \rrbracket_{\gamma}^{\mathrm{B}}=\left(\Psi, \gamma^{\prime}, \emptyset\right)$, except for

$$
\begin{gathered}
\frac{\llbracket e \rrbracket_{\gamma}^{\mathrm{EB}}=(\Psi, \vec{b}, \varrho)}{\llbracket x=e \rrbracket_{\gamma}^{\mathrm{EB}}=(\Psi, \gamma[x / \vec{b}], \varrho)} \\
\tau_{0} f\left(\tau_{1} x_{1}, \cdots, \tau_{m} x_{m}\right)\left\{S_{1} ; \cdots ; S_{n} ; \operatorname{return} y ;\right\} \gamma_{0}=\gamma\left[x_{1} / \gamma\left(y_{1}\right)\right] \cdots\left[x_{m} / \gamma\left(y_{m}\right)\right] \\
\llbracket S_{1} \rrbracket_{\gamma_{0}}^{\mathrm{EB}}=\left(\Psi_{1}, \gamma_{1}, \varrho_{1}\right), \cdots, \llbracket S_{n} \rrbracket_{\gamma_{n-1}}^{\mathrm{EB}}=\left(\Psi_{n}, \gamma_{n}, \varrho_{n}\right)
\end{gathered}
$$

Intuitively, the semantic rule $\llbracket x=g\left(y_{1}, \cdots, y_{m}\right) \rrbracket_{\gamma}^{\mathrm{EB}}$ sums up the expressions $\varrho_{i}$ 's of the statements $S_{i}$ 's in the function body because of $\prod_{i=1}^{n} 2^{-\varrho_{i}}=2^{-\sum_{i=1}^{n} \varrho_{i}}$.

### 7.3 Extended Bit-wise Resistance Evaluation

To evaluate the resistance of the program $P$ in an extended bit-wise manner, we define the (extended bit-wise) semantics $\llbracket P \rrbracket^{\text {EB }}$ of the program $P$ using its fn function $f$ as follows:

$$
\llbracket P \rrbracket^{\mathbb{E B}}=\llbracket \text { uints }[n] f\left(\text { uints }\left[n_{1}\right] k \text {, uints }[n] t x t\right)\left\{S_{1} ; \cdots ; S_{n} ; \text { return } y ;\right\} \rrbracket^{\mathrm{EB}}=\left(\Psi, \gamma_{n}, \varrho\right)
$$

where $\Psi=\bigcup_{i=1}^{n} \Psi_{i}, \varrho=\sum_{i=1}^{n} \varrho_{i}, \llbracket S_{1} \rrbracket_{\gamma_{0}}^{B}=\left(\Psi_{1}, \gamma_{1}, \varrho_{1}\right), \cdots, \llbracket S_{n} \rrbracket_{\gamma_{n-1}}^{\mathrm{B}}=\left(\Psi_{n}, \gamma_{n}, \varrho_{n}\right)$ and $\gamma_{0}$ is an initial state mapping each bit of array elements of $t x t$ to a fresh Boolean variable. We get that:

Theorem 7.2. Let $\llbracket P \rrbracket^{\text {EB }}=(\Phi, \gamma, \varrho)$ and $u$ be the minimum value of $\varrho$ subject to the set of IL constraints $\Phi \cup\left\{\sum_{i=0}^{s \cdot n-1} \gamma(t x t, i) \geq 1\right\}$. The MaxEDCP of the program $P$ is no greater than $2^{-u}$.

Similar to the decomposition approach given in Proposition 5.3, we have
Proposition 7.3. Let $u_{1}$ and $u_{2}$ be the MaxEDCP of the first $s_{1}$-round and the subsequent $s_{2}$ round differential characteristics. Then, $u_{1} \cdot u_{2}$ is an upper bound of the MaxEDCP of $\left(s_{1}+s_{2}\right)$-round differential characteristics.

By Theorem 7.2 and Proposition 7.3, we have the following corollaries.
Corollary 7.4. Let $\llbracket P \rrbracket^{\mathrm{EB}}=(\Phi, \gamma, \varrho)$. If $\{\varrho<\mathbb{C}\} \cup \Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$ is unsatisfiable, then the program $P$ with block size $d$ is resistant against differential cryptanalysis.

Corollary 7.5. If the s-round cipher $P$ can be partitioned into $\frac{s}{s^{\prime}}$ identical $s^{\prime}$-round ciphers $P^{\prime}$ such that $\llbracket P^{\prime} \rrbracket^{\mathrm{EB}}=(\Phi, \gamma, \varrho)$ and $\left\{\varrho<\frac{s^{\prime} \cdot \mathscr{\ell}}{s}\right\} \cup \Phi \cup\left\{\left(\sum_{i=0}^{n-1} \gamma(t x t, i)\right) \geq 1\right\}$ is unsatisfiable, then the program $P$ with block size $\&$ is resistant against differential cryptanalysis.

Table 3. Statistics of NIST candidates.

| Name | EasyBC |  |  | C/C++ |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\mathrm{LOC}_{\text {bw }}$ | $\mathrm{LOC}_{\text {ww }}$ | Time | LOC | Time |
| ASCON $\left(p^{a}\right)$ [Dobraunig et al. 2016]* | 55 | N/A | 27.7 | 42 | 1.4 |
| ELEPHANT [Beyne et al. 2020]* | 37 | N/A | 65.3 | 69 | 6.6 |
| GIFT-COFB-128 [Banik et al. 2020] | 34 | N/A | 23.9 | 81 | 3.0 |
| GRAIN(AEAD) [Hell et al. 2021]* | 99 | N/A | 0.1 | 146 | 0.1 |
| ISAP(v2.0) [Dobraunig et al. 2020]* | 72 | N/A | 141.8 | 94 | 0.0 |
| PHOTON(Beetle) [Bao et al. 2019]* | - | 51 | 24.4 | 104 | 1.8 |
| ROMULUS-128 [Iwata et al. 2020] | 57 | N/A | 32.3 | 113 | 0.2 |
| SPARKLE256(slim) [Beierle et al. 2019] ${ }^{*}$ | 39 | N/A | 0.3 | 19 | 0.0 |
| TinyJAMBU [Wu and Huang 2019]* | 15 | N/A | 143.3 | 18 | 0.5 |
| XOODYAK [Daemen et al. 2020]* | 38 | N/A | 21.1 | 63 | 0.1 |

Time is in milliseconds. N/A: word-wise is not applicable. -b gives the block size required for security evaluation of block ciphers. $\mathrm{LOC}_{\mathrm{bw}} / \mathrm{LOC}_{\mathrm{ww}}$ : LOC of bit-/word-wise implementation. *indicates keyless permutations and ${ }^{\star}$ indicates stream cipher where no block size is given.

Table 4. Statistics of other block ciphers.

| Name | EASYBC |  |  | C/C++ |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
|  | LOC $_{\text {bw }}$ LOC $_{\text {ww }}$ | Time | LOC | Time |  |
| AES-128 [Daemen and Rijmen 1999] | - | $\mathbf{7 1}$ | 9.0 | 106 | 12.5 |
| DES-64 [Fox 2000] | $\mathbf{5 0}$ | N/A | 4.0 | 77 | 2.4 |
| GIFT-64 [Banik et al. 2017] | $\mathbf{3 4}$ | N/A | 6.5 | 63 | 1.9 |
| KLEIN-64 [Gong et al. 2011] | - | $\mathbf{6 4}$ | 8.1 | 97 | 0.5 |
| LBLOCK-64 [Wu and Zhang 2011] | - | $\mathbf{4 4}$ | 6.0 | 78 | 1.7 |
| MIBS-64 [Izadi et al. 2009] | - | $\mathbf{3 7}$ | 13.9 | 69 | 0.3 |
| PICCOLO-64 [Shibutani et al. 2011] | - | $\mathbf{8 5}$ | 7.4 | 90 | 84.5 |
| PRESENT-64 [Bogdanov et al. 2007] | $\mathbf{2 8}$ | N/A | 8.1 | 87 | 0.6 |
| RECTANGLE-64 [Zhang et al. 2015] | $\mathbf{2 8}$ | N/A | 6.6 | 80 | 1.6 |
| SIMON-32 [Beaulieu et al. 2015] | $\mathbf{3 5}$ | N/A | 3.0 | 46 | 0.8 |
| SIMON-48 [Beaulieu et al. 2015] | $\mathbf{3 5}$ | N/A | 5.9 | 46 | 7.1 |
| SKINNY-64 [Beierle et al. 2016] | $\mathbf{3 8}$ | N/A | 10.7 | 67 | 0.3 |
| TWINE-64 [Suzaki et al. 2012] | - | $\mathbf{4 3}$ | 9.8 | 61 | 4.4 |

## 8 EVALUATION

Our approach is implemented as an open-source tool. As shown in Figure 5, it utilizes the SMT solver Z3 for computing the branch number and solving MaxSMT problems and Gurobi [Gurobi Optimization 2018] for solving MILP. In general, EASyBC iteratively increases the round number from 1 (cf. Corollary 5.4 and Corollary 7.4). It also partitions an $s$-round cipher to the maximum number of identical $s^{\prime}$-round ciphers and iteratively increases the round number $s^{\prime}$ (cf. Corollary 5.5 and Corollary 7.5), until the resistance is proved. The tool is designed to be modular and extensible, where each semantic rule has an API wrapper and alternative generation methods can be easily chosen and added. We also incorporate the bounding condition [Matsui 1994; Zhang et al. 2018] into MILP to prune search space which often improves the overall MILP solving. (The detail is given in [Sun et al. 2023, Section E.4].)

All the experiments were conducted on a machine with two Intel Xeon Gold 5118 CPUs (12 cores, 2.30 GHz ), 64 -bit Ubuntu 20.04 LTS, and 128 GB RAM. The number of threads for Gurobi is set to 16.

### 8.1 Expressiveness of EASYBC

To evaluate the expressiveness of EAsyBC, we implement 23 realistic cryptographic primitives with EASYBC, consisting of all the 10 finalists of the NIST lightweight cryptography standardization process [NIST 2023] and 13 commonly used block ciphers, covering both SPN ciphers (e.g., AES, PRESENT, and GIFT-COFB) and BFN ciphers (e.g., DES, LBLOCK, TWINE). We stress that we focus on the underlying primitives (i.e., block ciphers and key-less permutations) rather than the full authenticated encryption, message authentication, or hash protocols in the NIST finalists.

The physical source lines of code (LOC [Nguyen et al. 2007]) counted by cloc [Danial 2021] are reported in Tables 3 and 4. We report LOC of word-wise (when available, or otherwise bit-wise) EAsyBC implementations. As a comparison, we also report LOC of the C/C++ reference implementations of the NIST finalists and (randomly selected) GitHub open-source C/C++ implementations of other block ciphers. To some extent, a smaller number (highlighted in bold) indicates that the language is more succinct in implementing cryptographic primitives.

We observe that EAsyBC is sufficiently expressive to easily implement all the 23 cryptographic primitives either in word-wise or bit-wise fashion. More specifically, when cryptographic primitives can be implemented in a word-wise fashion (i.e., PHOTON, AES, KLEIN, LBLOCK, MIBS, PICCOLO, and TWINE), their EAsyBC implementations always require considerably less code than the baselines. The main reason is that EAsyBC provides high-level constructs of P-box and matrix-vector products for permutations and linear transformations. For cryptographic primitives

Table 5. Results of the word-wise approach, where ( $n$ ) indicates the number of entire rounds of the cipher and \#AS denotes the lower bound of the minimum number of active S-boxes.

| Rounds |  | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 | 17 | 18 | 19 | 20 | 21 | 22 | 23 | 24 | 25 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| AES (14) | \#AS | 1 | 5 | 9 | 25 | 26 | 30 | 34 | 50 | 51 | 55 | 59 | 75 | 76 | 80 | N/A |  |  |  |  |  |  |  |  |  |  |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 1 s | 0s | 0s | 0s | 1 s | 0s | 0s | 0s | N/A |  |  |  |  |  |  |  |  |  |  |
| KLEIN (12) | \#AS | 1 | 5 | 8 | 15 | 16 | 20 | 23 | 30 | 31 | 35 | 38 | 45 | N/A |  |  |  |  |  |  |  |  |  |  |  |  |
|  | Time | 1 s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | N/A |  |  |  |  |  |  |  |  |  |  |  |  |
| LBLOCK (32) | \#AS | 0 | 1 | 2 | 3 | 4 | 6 | 8 | 11 | 14 | 18 | 22 | 24 | 27 | 30 | 32 | 35 | 36 | 39 | 41 | 44 | 45 | 48 | 50 | 53 | 54 |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 1s | 0s | 1s | 1s | 1s | 3 s | 3s | 2s | 6 s | 5s | 62s | 62s | 80s |
| MIBS (32) | \#AS | 0 | 1 | 2 | 5 | 6 | 7 | 8 | 11 | 12 | 13 | 14 | 17 | 18 | 19 | 20 | 23 | 24 | 25 | 26 | 29 | 30 | 31 | 32 | 35 | 36 |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 1s | 0s | 0s | 1s | 1s | 1s | 3s | 4 s | 2s | 4s | 2s | 3s | 2s | 3 s | 4s | 5s | 4 s | 7s |
| PHOTON (12) | \#AS | 1 | 9 | 17 | 81 | 82 | 90 | 98 | 162 | 163 | 171 | 179 | 243 | N/A |  |  |  |  |  |  |  |  |  |  |  |  |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 1s | 0s | N/A |  |  |  |  |  |  |  |  |  |  |  |  |
| PICCOLO (25) | \#AS | 0 | 5 | 10 | 15 | 20 | 30 | 35 | 40 | 45 | 50 | 55 | 60 | 65 | 70 | 75 | 80 | 85 | 90 | 95 | 100 | 105 | 110 | 115 | 120 | 125 |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 1s | 0s | 1s | 0s | 1 s | 3s | 6 s | 4s | 6s | 28s | 30s | 31s | 38s | 187s | 93s | 122s | 289s | 560s |
| TWINE (36) | \#AS | 0 | 1 | 2 | 3 | 4 | 6 | 8 | 11 | 14 | 18 | 22 | 24 | 27 | 30 | 32 | 35 | 36 | 39 | 41 | 44 | 45 | 48 | 50 | 53 | 54 |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 0s | 1 s | 0 s | 0s | 0s | 1 s | 0s | 1 s | 2 s | 2 s | 3 s | 2s | 5 s | 6 s | 26s | 79 s | 72 s |

that are implemented in a bit-wise fashion with EASYBC, their EASYBC implementations also require significantly less code than their baselines except for ASCON and SPARKLE, due to the following reasons. (1) One 320 -bit block is stored in five 64 -bit variables in the reference implementation of ASCON each of which is permuted, while one 320 -bit block is stored in one Boolean array in the EASYBC implementation so that the 320 -bit Boolean array has to be split before the permutation and merged after permutation. (2) The C++ reference implementation of SPARKLE uses function-like macro definitions and thus is more succinct.
Results of EAsYBC interpreter. For each realistic cryptographic primitive, we randomly generate 100 inputs to EAsyBC programs and binary executables of C/C++ programs. We run the EAsyBC program (using our interpreter) and the binary executable for each input and record the output and the execution time. The outputs of the respective programs have been compared with $100 \%$ match, which validates the semantics and the interpreter of EAsYBC, as well as EAsyBC programs. The average execution times over 100 inputs are reported in Tables 3 and 4. While executing via our interpreter is less efficient, it is acceptable for testing EAsyBC programs.

### 8.2 Effectiveness of EASyBC

To evaluate the effectiveness of EASyBC, we first compare the performance of various alternative methods to generate MILP. According to our experiment results (cf. [Sun et al. 2023, Section E]), we select the optimal modeling methods for cryptographic primitives, i.e., $\Psi_{2,3}^{1}$ and $\Psi_{M}^{1}$ are used for modeling the modular addition, substitution, XOR and matrix-vector product respectively in the word-wise approach (cf. Figure 6); $\psi_{\oplus}^{2}, \Psi_{M, i, h}^{2}$ and the technique of [Boura and Coggia 2020, Alg. 2, Alg. 3 and Proposition 3] are used for modeling XOR, matrix-vector product and constructing $\Psi_{\mathcal{S}}^{3}$ in the bit-wise approach (cf. Figure 8 and Section 4.2); the technique of [Sasaki and Todo 2017] is used for constructing $\Psi_{\mathcal{S}}^{4}$ in the extended bit-wise approach (cf. Section 4.3). We remark that there is no consensus on the security of key-less permutations yet, and they are used to evaluate the performance of EAsYBC instead of security evaluation.
8.2.1 Word-wise Approach. The word-wise approach is evaluated on all the word-wise implementations. The results are reported in Table 5 up to 25 rounds which suffice to prove security. The results for the entire rounds are given in [Sun et al. 2023, Section E.1].

We observe that the execution time (i.e., the MILP solving time) in seconds (s) increases with the round number. The execution time of PICCOLO is considerably higher than that of the others when the round number is large (e.g., $\geq 13$ ), because PICCOLO generates more constraints. For instance, the number of constraints for the 20 -round LBLOCK and TWINE are both 481, while it is

Table 6. Results of the bit-wise approach with the bounding condition, where Timeout is 24 hours.

| Rounds |  | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ASCON (12) | \#AS 1 |  | 4 | 15 | N/A |  |  |  |  |  |  |  |  |  |  |  |
|  | Time | 0s | 1061s | 1573s | Timeout |  |  |  |  |  |  |  |  |  |  |  |
| DES (16) | \#AS 0 |  | 1 | 2 | 4 | 6 | 8 | 9 | 12 | 12 | N/A |  |  |  |  |  |
|  | Time | 0s | 0s | 0s | 6 s | 61s | 355s | 1073s | 27294s | 57344s | Timeout |  |  |  |  |  |
| ELEPHANT (80) | \#AS | 1 | 2 | 4 | 6 | 10 | 12 | 14 | 16 | 18 | 20 | 22 | 24 | N/A |  |  |
|  | Time | 0s | 0s | 4s | 5s | 50s | 394s | 1653s | 1592s | 2994s | 4353s | 18294s | 20404s | Timeout |  |  |
| GIFT-COFB (40) | \#AS |  | 2 | 3 | 5 | 7 | 10 | 13 | 17 | 19 | 21 | N/A |  |  |  |  |
|  | Time | 0s | 0s | 1s | 7 s | 7s | 38s | 775s | 2222s | 6725s | 77870s | Timeout |  |  |  |  |
| GIFT (28) | \#AS | 1 | 2 | 3 | 5 | 7 | 10 | 13 | 16 | 18 | 20 | 22 | 24 | 26 | N/A |  |
|  | Time | 0s | 0s | 1s | 1s | 2s | 3s | 7 s | 52s | 43s | 136s | 518s | 846 s | 25561s | Timeout |  |
| PRESENT (31) | \#AS |  | 2 | 4 | 6 | 10 | 12 | 14 | 16 | 18 | 20 | 22 | 24 | 26 | 28 | N/A <br> Timeout |
|  | Time | 0s | 0s |  | 1s | 6 s | 8 s | 15s | 65s | 95s | 539s | 1884s | 10271s | 38907s | 50931s |  |
| RECTANGLE (25) | \#AS |  | 2 | 3 | 4 | 6 | 8 | 11 | 13 | 15 | 17 | 19 | 21 | 23 | 25 | 27 <br> 37860 s |
|  | Time | 1s | 0s | 0s | 1s | 8 s | 21s | 6 s | 436s | 63s | 603s | 1135s | 1221s | 2841s | 36333s |  |
| ROMULUS (40) | \#AS 1 |  | 2 | 5 | 8 | 12 | 16 | 26 | 36 | 41 | N/A |  |  |  |  |  |
|  | Time | 0s | 0s |  | 48s | 87s | 378s | 1526s | 17266s | 4043s | Timeout |  |  |  |  |  |
| SKINNY (36) | \#AS |  | 2 | 1s | 8 | 12 | 16 | 26 | 36 | 41 | 46 | 51 | 55 | N/A |  |  |
|  | Time | 0s | 0s |  | 2s | 15s | 24s | 78s | 967 s | 2041s | 3610s | 15502s | 26989s | Timeout |  |  |
| SPARKLE (7) | \#AS |  | 2 | 1624s | 6 | 7 | N/A |  |  |  |  |  |  |  |  |  |
|  | Time | 2s | 44s |  | 4416s | 17592s | Timeout |  |  |  |  |  |  |  |  |  |

641 for the 20 -round PICCOLO. However, the large round number is not always necessary, as the security may have been proved with a small round number (see below).

For block ciphers, $p=2^{-6}$ and $a=128$ for AES; $p=2^{-2}$ and $b=64$ for KLEIN, LBLOCK, MIBS, PICCOLO and TWINE; $p=2^{-2}$ and $a=128$ for PHOTON, where the maximum probability $p$ of the involved S-boxes that are arrays is computed by enumeration. By Corollary 5.4, EASyBC proved that AES (resp. KLEIN, LBLOCK, MIBS, PHOTON, PICCOLO and TWINE) is resistant when the round number $s$ is 4 (resp. 10, 15, 23, 4, 7 and 15), as highlighted in boldface in Table 5.
8.2.2 Bit-wise Approach. The bit-wise approach is evaluated on all bit-wise implementations with S-boxes. (The word-wise implementations are excluded.) The results are given in Table 6 up to 15 rounds within 24 hours, where the execution time is the MILP solving time.

Unsurprisingly, we observe that the execution time increases very quickly with the round number due to the blow-up of constraints. For instance, the numbers of constraints and involved variables of ASCON are 6,913 and 1,408 respectively for 1 round, but become 13,825 and 2,496 (resp. 20,737 and 3,584 ) for 2 (resp. 3) rounds.

For block ciphers, $p=2^{-2}$ and $b=64$ for DES, PRESENT, RECTANGLE and SKINNY; $p=2^{-1.415}$ and $a=128$ for GIFT-COFB; $p=2^{-1.415}$ and $a=64$ for GIFT; and $p=2^{-2}$ and $b=128$ for ROMULUS. We found that by Corollary 5.4, EAsyBC cannot prove that DES (resp. GIFT-COFB, GIFT, PRESENT, RECTANGLE, ROMULUS and SKINNY) is resistant using the lower bounds of numbers of active S-boxes reported in Table 6. Fortunately, by Corollary 5.5, EasyBC proved that GIFT (resp. PRESENT, RECTANGLE, ROMULUS and SKINNY) is resistant with 6 (resp. 3, 6, 3 and 1) rounds, as highlighted in boldface in Table 6. Note that the resistance of GIFT-COFB cannot be proved here which will be done by applying the extended bit-wise approach later, while DES is indeed vulnerable to differential cryptanalysis [Biham and Shamir 1990] and EASYBC can return the differential characteristics up to 9 rounds within 24 -hour time limit.

We also compared the MILP solving time with/without bounding conditions. Adding the bounding condition often (5 out of 8) improves the efficiency by 1 to 3 times, but does not necessarily improve (and sometimes even worsens) the efficiency (3 out of 8). The results without the bounding condition are given in [Sun et al. 2023, Section E.5].
8.2.3 Extended Bit-wise Approach. The capability of the extended bit-wise approach is evaluated on all the bit-wise implementations of block ciphers and the S-box of SPARKLE (i.e., 4-round Alzette)

Table 7. Results of the extended bit-wise approach with the bounding condition, where Pr denotes the upper bound of the probability of optimal differential characteristics, and Timeout is 24 hours.

| Rounds $r$ |  | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| GIFT-COFB (40) | Pr | $2^{-1.415}$ | $2^{-3.415}$ | $2^{-7}$ | $2^{-11.415}$ | $2^{-17}$ | $2^{-22.415}$ | $2^{-28.415}$ | N/A |  |  |  |  |  |  |  |
|  | Time | 0s | 6s | 9s | 124s | 1297s | 5811s | 11538s | Timeout |  |  |  |  |  |  |  |
| SIMON-32 (32) | Pr | $2^{0}$ | $2^{-2}$ | $2^{-4}$ | $2^{-6}$ | $2^{-8}$ | $2^{-12}$ | $2^{-14}$ | $2^{-18}$ | $2^{-20}$ | $2^{-26}$ | $2^{-30}$ | $2^{-34}$ | $2^{-36}$ | $2^{-38}$ | $2^{-40}$ |
|  | Time | 0s | 0s | 0s | 0s | 0s | 0s | 1s | 4s | 3s | 7s | 84s | 57s | 820s | 191s | 6327s |
| SIMON-48 (36) | Pr | $2^{0}$ | $2^{-2}$ | $2^{-4}$ | $2^{-6}$ | $2^{-8}$ | $2^{-12}$ | $2^{-14}$ | $2^{-18}$ | $2^{-20}$ | $2^{-26}$ | $2^{-30}$ | $2^{-36}$ | $2^{-38}$ | $2^{-44}$ | $2^{-46}$ |
|  | Time | 0s | 0s | 1s | 0s | 0s | 1 s | 3 s | 5 s | 14s | 16s | 101s | 95s | 402s | 505s | 53920s |
| Alzette (12) | Pr | $2^{0}$ | $2^{-1}$ | $2^{-2}$ | $2^{-6}$ | $2^{-10}$ | $2^{-18}$ | N/A |  |  |  |  |  |  |  |  |
|  | Time | 0s | 1 s | 2 s | 75s | 221s | 3897s | Timeout |  |  |  |  |  |  |  |  |

that cannot be proved or analyzed before. The results are given in Table 7 up to 15 rounds within 24 hours, where the execution time is the MILP solving time.

Unsurprisingly, the execution time of the extended bit-wise approach is longer than that of the bit-wise approach. For instance, on the 7-round GIFT-COFB, the execution time of the extended bit-wise approach is $11,538 \mathrm{~s}$ while the execution time of the bit-wise approach is $1,074 \mathrm{~s}$. This is because probabilities are explicitly encoded using additional Boolean variables in the extended bit-wise approach, resulting in more difficult MILP instances.

Note the block size: GIFT-COFB $b=128$, SIMON-32 $b=32$, and SIMON- $48 a=48$. By Corollary 7.5, EAsyBC prove that GIFT-COFB (resp. SIMON-32 and SIMON-48) is resistant with 5 (resp. 2 and 3) rounds.

## 9 RELATED WORK

[Matsui 1994] proposed the first algorithm for automated resistance analysis of block ciphers against differential cryptanalysis. To further enhance efficiency, various heuristics have been proposed to reduce the search space [Aoki et al. 1997; Bao et al. 2014; Biryukov and Nikolić 2010; Ji et al. 2021]. However, these heuristics generally rely on cipher-specific optimizations, necessitating sophisticated programming skills. Additionally, creating highly reusable code that can be easily adapted for different ciphers is non-trivial [Zhang et al. 2018].

In recent years, a more promising approach based on MILP has been developed. [Mouha et al. 2011] proposed to determine the lower bound of the minimum number of active S-boxes via MILP solving. As an early attempt, it only considered the word-wise modeling for the XOR operation, S-box, and linear transformation. To partially lift this limitation, [Sun et al. 2013] introduced the bitwise modeling. To precisely characterize S-boxes in MILP, [Sun et al. 2014a] proposed to construct IL constraints from the H-representation of the S-box DDT and reduced the number of constraints via a greedy algorithm. Later, Sun et al. [Sun et al. 2014b] proposed a bit-wise modeling method for the AND operation and extended the method of [Sun et al. 2014a] to directly bound the MaxEDCP by encoding the probabilities between input and output differences of S-boxes in IL constraints.

Since then, plenty of modeling methods for specific operations have been proposed, aimed at improving efficiency and applicability. [Sasaki and Todo 2017] proposed an MILP-based algorithm to reduce the number of constraints obtained from the H-representation of the S-box DDT. [Abdelkhalek et al. 2017] proposed to construct IL constraints from the minimized product-of-sum representation of S-box DDT instead of the H-representation. Along this line, [Boura and Coggia 2020; Li and Sun 2022; Sun 2021; Udovenko 2021] generate more diverse constraints from the S-box DDT, allowing the number of the resulting constraints to be further reduced by applying greedy or MILP-based algorithms; [Cui et al. 2016; Li et al. 2019; Sasaki and Todo 2017; Yin et al. 2017] proposed a new modeling for the XOR operation; [Boura and Coggia 2020; Ilter and Selçuk 2021; Zhang and Zhang 2018] proposed another modeling for linear transformation; [Fu et al. 2016] proposed a modeling method for the modular addition operation; and [Chen et al. 2015; Liu et al.

2017; Wang et al. 2018] proposed a modeling method for the rotation-AND operation. Besides new modeling methods for specific operations, optimization strategies have also been proposed. [Zhang et al. 2018] incorporated the bounding condition of [Matsui 1994] into the MILP-based method; [Zhou et al. 2019] proposed partitioning all possible differential characteristics into subsets, each of which is analyzed by the MILP-based method.

Another direction is to resort to SAT/SMT solving. [Mouha and Preneel 2013] proposed the first SAT/SMT modeling for the XOR, modular addition, and rotation operations. It has been extended to handle a specific permutation [Aumasson et al. 2014], the AND operation and rotation with constants [Kölbl et al. 2015], independent modular addition [Song et al. 2016], S-boxes [Liu et al. 2021; Sun et al. 2018], and modular addition with constants [Azimi et al. 2022]. In contrast to the MILP-based method which determines the lower bound of the minimum number of active S-boxes or the upper bound of MaxEDCP, SAT/SMT-based methods can only verify whether a given number is a bound. Recently, both the bounding condition of [Matsui 1994] and MILP-based method have been combined with the SAT/SMT-based method [Makarim and Rohit 2022; Sun et al. 2021] to improve the efficiency. To facilitate the comparison of all the above MILP/SAT/SMT-based approaches, a summary is given in [Sun et al. 2023, Section F].

Despite significant progress in the field, existing works primarily concentrate on ad-hoc modeling methods designed for specific operations, but do not offer systematic methods for determining word-wise or bit-wise branch numbers and encoding probabilities between input and output differences. There is a lack of language support, unified computational approaches and full automation. This limitation forces cryptanalysts to individually model each cipher within the tool or create a model generation script for every individual cipher, resulting in a complex, error-prone, and time-consuming process. This work fills this significant gap and makes resistance evaluation against differential cryptanalysis easily accessible to cryptographers.

There are other cryptography-specific languages such as SAW [Carter et al. 2013], Jasmin [Almeida et al. 2017], Vale [Bond et al. 2017], Usuba [Mercadier and Dagand 2019], FaCT [Cauligi et al. 2019], QMVerif [Gao et al. 2022], HOME [Gao et al. 2021], and FISCHER [Liu et al. 2023]. They are designed to ensure functional correctness and/or side-channel security of cryptographic algorithms or implementations, which are considerably different from EasyBC.

## 10 CONCLUSION

We have designed a high-level cryptography-specific language EAsyBC for describing block ciphers and presented a rigorous MILP generation procedure from EASYBC programs in the form of differential denotational semantics, leading to a generic and extensible approach for automatically evaluating the resistance of block ciphers written in EAsyBC against differential cryptanalysis. We have implemented our approach in an open-source tool and extensively evaluate it on a set of realistic cryptographic primitives, demonstrating its expressivity and capability. In particular, experimental results show that realistic cryptographic primitives can be easily described in EASYBC and their resistance against differential cryptanalysis can be efficiently and effectively proved using our tool. Our tool enables cryptanalysts to easily assess the resistance of block ciphers against differential cryptanalysis in a fully automatic way.

For future research, it would be interesting to improve efficiency by combining recent optimization strategies and to develop analysis approaches for other powerful cryptanalysis (e.g., linear cryptanalysis [Biryukov and De Cannière 2011], impossible differential cryptanalysis [Kim et al. 2003]) based on EAsyBC.

## ACKNOWLEDGEMENT

We thank the reviewers of POPL'24 for their constructive and insightful comments. This work is supported by the National Natural Science Foundation of China (NSFC) under Grants No. 62072309 and No. 61872340, CAS Project for Young Scientists in Basic Research (YSBR-040), ISCAS New Cultivation Project (ISCAS-PYFX-202201), overseas grants from the State Key Laboratory of Novel Software Technology, Nanjing University (KFKT2023A04), and Birkbeck BEI School Project (EFFECT).

## DATA AVAILABILITY STATEMENT

The full version of the paper [Sun et al. 2023] contains missing proofs and more experimental results. Our tool is available at https://github.com/S3L-official/EasyBC.

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Received 2023-07-11; accepted 2023-11-07


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