# **Linearization Framework for Collision Attacks:** Application to CubeHash and MD6 (Extended Version\*)

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**Abstract.** In this paper, an improved differential cryptanalysis framework for finding collisions in hash functions is provided. Its principle is based on linearization of compression functions in order to find low weight differential characteristics as initiated by Chabaud and Joux. This is formalized and refined however in several ways: for the problem of finding a conforming message pair whose differential trail follows a linear trail, a condition function is introduced so that finding a collision is equivalent to finding a preimage of the zero vector under the condition function. Then, the dependency table concept shows how much influence every input bit of the condition function has on each output bit. Careful analysis of the dependency table reveals degrees of freedom that can be exploited in accelerated preimage reconstruction under the condition function. These concepts are applied to an in-depth collision analysis of reduced-round versions of the two SHA-3 candidates CubeHash and MD6, and are demonstrated to give by far the best currently known collision attacks on these SHA-3 candidates.

Key words: Hash functions, collisions, differential attack, SHA-3, CubeHash and MD6.

### Introduction

Hash functions are important cryptographic primitives that find applications in many areas including digital signatures and commitment schemes. A hash function is a transformation which maps a variable-length input to a fixed-size output, called message digest. One expects a hash function to possess several security properties, one of which is collision resistance. Being collision resistant, informally means that it is hard to find two distinct inputs which map to the same output value. In practice, the hash functions are mostly built from a fixed input size compression function, e.g. the renowned Merkle-Damgård construction. To any hash function, no matter how it has been designed, we can always attribute fixed input size compression functions, such that a collision for a derived compression function results in a direct collision for the hash function itself.

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This way, firstly we are working with fixed input size compression functions rather than varying input size ones, secondly we can attribute compression functions to those hash functions which are not explicitly based on a fixed input size compression function, and thirdly we can derive different compression functions from a hash function. For example multi-block collision attack [27] benefits from the third point. Our task is to find two messages for an attributed compression function such that their digests are preferably equal (a collision) or differ in only a few bits (a near-collision). Collisions for a compression function are directly translated into collisions for the hash function. The relevance of near collisions, however, depends on the hash function structure and the way we have defined the compression function. In most of the cases the near collisions of the compression function provide near collisions for the underlying hash function as well, but in some cases, like in sponge constructions with a strong filtering at the end or a Merkle-Damgård construction with a strong final transformation, they are of little interest.

The goal of this work is to *revisit* collision-finding methods using linearization of the compression function in order to find differential characteristics for the compression function. This method was initiated by Chabaud and Joux on SHA–0 [11] and was later extended and applied to SHA–1 by Rijmen and Oswald [26]. The recent attack on EnRUPT by Indesteege and Preneel [15] is another application of the method. In particular, in [26] it was observed that the codewords of a linear code, which are defined through a linearized version of the compression function, can be used to identify differential paths leading to a collision for the compression function itself. This method was later extended by Pramstaller et al. [25] with the general conclusion that finding high probability differential paths is related to low weight codewords of the attributed linear code. In this paper we further investigate this issue.

The first contribution of our work is to present a more concrete and tangible relation between the linearization and differential paths. In the case that modular addition is the only involved nonlinear operation, our results can be stated as follows. Given the parity check matrix  $\mathcal{H}$  of a linear code, and two matrices  $\mathcal{A}$  and  $\mathcal{B}$ , find a codeword  $\Delta$  such that  $\mathcal{A}\Delta \vee \mathcal{B}\Delta$  is of low weight. This is clearly different from the problem of finding a low weight codeword  $\Delta$ . We then consider the problem of finding a conforming message pair for a given differential trail for a certain linear approximation of the compression function. We show that the problem of finding conforming pairs can be reformulated as finding preimages of zero under a function which we call the *condition* function. We then define the concept of *dependency table* which shows how much influence every input bit of the condition function has on each output bit. By carefully analyzing the dependency table, we are able to profit not only from neutral bits [7] but also from probabilistic neutral bits [2] in a backtracking search algorithm, similar to [6, 14, 24]. This contributes to a better understanding of freedom degrees uses.

We consider compression functions working with n-bit words. In particular, we focus on those using modular addition of n-bit words as the only nonlinear operation. The incorporated linear operations are XOR, shift and rotation of n-bit words in practice. We present our framework in detail for these constructions by approximating modular addition with XOR. We demonstrate its validity by applying it on reduced-round variants of CubeHash [4] (one of the NIST SHA-3 [22] competitors) which uses addition,

XOR and rotation. CubeHash instances are parametrized by two parameters r and b and are denoted by CubeHash-r/b which process b message bytes per iteration; each iteration is composed of r rounds. Although we can not break the original submission CubeHash-8/1, we provide real collisions for the much weaker variants CubeHash-3/64 and CubeHash-4/48. Interestingly, we show that neither the more secure variants CubeHash-6/16 and CubeHash-7/64 do provide the desired collision security for 512-bit digests by providing theoretical attacks with complexities  $2^{222.6}$  and  $2^{203.0}$  respectively; nor that CubeHash-6/4 with 512-bit digests is second-preimage resistant, as with probability  $2^{-478}$  a second preimage can be produced by only one hash evaluation. Our theory can be easily generalized to arbitrary nonlinear operations. We discuss this issue and as an application we provide collision attacks on 16 rounds of MD6 [23]. MD6 is another SHA-3 candidate whose original number of rounds varies from 80 to 168 when the digest size ranges from 160 to 512 bits.

# 2 Linear differential cryptanalysis

Let's consider a compression function  $H = \mathsf{Compress}(M, V)$  which works with n-bit words and maps an m-bit message M and a v-bit initial value V into an h-bit output H. Our aim is to find a collision for such compression functions with a randomly given initial value V. In this section we consider modular-addition-based Compress functions, that is, they use only modular additions in addition to linear transformations. This includes the family of AXR (Addition-XOR-Rotation) hash functions which are based on these three operations. In Section 5 we generalize our framework to other family of compression functions. For these Compress functions, we are looking for two messages with a difference  $\Delta$  that result in a collision. In particular we are interested in a  $\Delta$  for which two randomly chosen messages with this difference lead to a collision with a high probability for a randomly chosen initial value. For modular-addition-based Compress functions, we consider a linearized version for which all additions are replaced by XOR. This is a common linear approximation of addition. Other possible linear approximations of modular addition, which are less addressed in literature, can be considered according to our generalization of Section 5. As addition was the only nonlinear operation, we now have a linear function which we call  $\mathsf{Compress}_{\mathsf{lin}}$ . Since  $\mathsf{Compress}_{\mathsf{lin}}(M,V) \oplus$  $\mathsf{Compress}_{\mathsf{lin}}(M \oplus \Delta, V) = \mathsf{Compress}_{\mathsf{lin}}(\Delta, 0)$  is independent of the value of V, we adopt the notation  $Compress_{lin}(M) = Compress_{lin}(M,0)$  instead. Let  $\Delta$  be an element of the kernel of the linearized compression function, i.e. Compress<sub>lin</sub>( $\Delta$ ) = 0. We are interested in the probability  $\Pr\{\mathsf{Compress}(M,V) \oplus \mathsf{Compress}(M \oplus \Delta,V) = 0\}$ for a random M and V. In the following we present an algorithm which computes this probability, called the raw (or bulk) probability.

### 2.1 Computing the raw probability

We consider a general n-bit vector  $x = (x_0, \ldots, x_{n-1})$  as an n-bit integer denoted by the same variable, *i.e.*  $x = \sum_{i=0}^{n-1} x_i 2^i$ . The Hamming weight of a binary vector or an integer x,  $\operatorname{wt}(x)$ , is the number of its nonzero elements, *i.e.*  $\operatorname{wt}(x) = \sum_{i=0}^{n-1} x_i$ . We use + for modular addition of words and  $\oplus$ ,  $\vee$  and  $\wedge$  for bit-wise XOR, OR and AND

logical operations between words as well as vectors. We use the following lemma which is a special case of the problem of computing  $\Pr\{\left((A \oplus \alpha) + (B \oplus \beta)\right) \oplus (A+B) = \gamma\}$  where  $\alpha, \beta$  and  $\gamma$  are constants and A and B are independent and uniform random variables, all of them being n-bit words. Lipmaa and Moriai have presented an efficient algorithm for computing this probability [19]. We are interested in the case  $\gamma = \alpha \oplus \beta$  for which the desired probability has a simple closed form.

**Lemma 1.** 
$$\Pr\{((A \oplus \alpha) + (B \oplus \beta)) \oplus (A + B) = \alpha \oplus \beta\} = 2^{-\operatorname{wt}((\alpha \vee \beta) \wedge (2^{n-1} - 1))}.$$

Lemma 1 gives us the probability that modular addition behaves like the XOR operation. As  $\mathsf{Compress}_{\mathsf{lin}}$  approximates  $\mathsf{Compress}$  by replacing modular addition with XOR, we can then devise a simple algorithm to compute (estimate) the raw probability  $\mathsf{Pr}\{\mathsf{Compress}(M,V)\oplus\mathsf{Compress}(M\oplus\Delta,V)=\mathsf{Compress}_{\mathsf{lin}}(\Delta)\}$ . Let's first introduce some notations.

**Notations.** Let  $n_{\mathrm{add}}$  denote the number of additions which Compress uses in total. In the course of evaluation of  $\mathsf{Compress}(M,V)$ , let the two addends of the i-th addition  $(1 \leq i \leq n_{\mathrm{add}})$  be denoted by  $A^i(M,V)$  and  $B^i(M,V)$ , for which the ordering is not important. The value  $C^i(M,V) = \left(A^i(M,V) + B^i(M,V)\right) \oplus A^i(M,V) \oplus B^i(M,V)$  is then called the carry word of the i-th addition. Similarly, in the course of evaluation of  $\mathsf{Compress}_{\mathrm{lin}}(\Delta)$ , denote the two inputs of the i-th linearized addition by  $\alpha^i(\Delta)$  and  $\beta^i(\Delta)$  in which the ordering is the same as that for  $A^i$  and  $B^i$ . We define five more functions  $\mathbf{A}(M,V)$ ,  $\mathbf{B}(M,V)$ ,  $\mathbf{C}(M,V)$ ,  $\alpha(\Delta)$  and  $\beta(\Delta)$  with  $(n-1)n_{\mathrm{add}}$ -bit outputs. These functions are defined as the concatenation of all the  $n_{\mathrm{add}}$  relevant words excluding their MSBs. For example  $\mathbf{A}(M,V)$  and  $\alpha(\Delta)$  are respectively the concatenation of the  $n_{\mathrm{add}}$  words  $\left(A^1(M,V),\ldots,A^{n_{\mathrm{add}}}(M,V)\right)$  and  $\left(\alpha^1(\Delta),\ldots,\alpha^{n_{\mathrm{add}}}(\Delta)\right)$  excluding the MSBs.

Table 1 is a reference for all the main notations used in sections 2, 3.1 and 5. Using these notations, the raw probability can be simply estimated as follows.

**Lemma 2.** Let Compress be a modular-addition-based compression function. Then for any message difference  $\Delta$  and for random values M and V,  $p_{\Delta} = 2^{-\operatorname{wt}\left(\alpha(\Delta)\vee\beta(\Delta)\right)}$  is a lower bound for  $\operatorname{Pr}\left\{\operatorname{Compress}(M,V) \oplus \operatorname{Compress}(M \oplus \Delta,V) = \operatorname{Compress}_{\operatorname{lin}}(\Delta)\right\}$ .

*Proof.* We start with the following definition.

**Definition 1.** We say that a message M (for a given V) conforms to (or follows) the trail of  $\Delta$  iff  $^4$ 

$$((A^i \oplus \alpha^i) + (B^i \oplus \beta^i)) \oplus (A^i + B^i) = \alpha^i \oplus \beta^i, \text{ for } 1 \le i \le n_{\text{add}},$$
 (1)

where  $A^i$ ,  $B^i$ ,  $\alpha^i$  and  $\beta^i$  are shortened forms for  $A^i(M,V)$ ,  $B^i(M,V)$ ,  $\alpha^i(\Delta)$  and  $\beta^i(\Delta)$ , respectively.

<sup>&</sup>lt;sup>4</sup> if and only if.

**Table 1.** The main notations used in sections 2, 3.1 and 5 along with their description.

	decentration
notation  H = Compress (M, V)	description
H = Compress(M, V)	a modular-addition-based or binary-FSM-based compres-
(4.0) 6 (4)	sion function from $\{0,1\}^m \times \{0,1\}^v$ to $\{0,1\}^h$
Compress $_{lin}(\Delta, 0)$ or Compress $_{lin}(\Delta)$	the linearized version of Compress
Condition $_{\Delta}(M, V)$ or Condition $(M, V)$	condition function associated with a differential trail $\Delta$ for
	a compression function Compress and it is linearized ver-
	sion Compress <sub>lin</sub>
Y	the output of the condition function, $Y = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$
V	Condition $\Delta(M, V)$
$Y_j$	the <i>j</i> -th condition bit
y	the total number of condition bits, <i>i.e.</i> length of Y
$n_{ m add}$	the total number of modular additions of a modular-
	addition-based Compress function
$n_{ m nl}$	the total number of NUBFs of a binary-FSM-based
(41 71) (41(15 17) 41(15 17))	Compress function
$(A^i, B^i)$ or $(A^i(M, V), A^i(M, V))$	the two addend words of <i>i</i> -th addition in a modular-
-d -d	addition-based Compress function
$C^i$ or $C^i(M,V)$	the carry word of the <i>i</i> -th addition in a modular-addition-
	based Compress function
$(\alpha^i, \beta^i)$ or $(\alpha^i(\Delta), \beta^i(\Delta))$	the two input words of the $i$ -th linearized modular addition
	(i.e. XOR) in a modular-addition-based linearized compres-
	sion function Compress <sub>lin</sub>
$\mathbf{A}(M,V)$	the concatenation of all $n_{\rm add}$ addend words
	$A^1, \ldots, A^{n_{\text{add}}}$ excluding their MSBs
$\mathbf{B}(M,V)$	the concatenation of all $n_{\text{add}}$ addend words
	$B^1, \dots, B^{n_{\text{add}}}$ excluding their MSBs
$\mathbf{C}(M,V)$	the concatenation of all $n_{\text{add}}$ carry words
(4)	$C^1, \dots, C^{n_{\text{add}}}$ excluding their MSBs
$\alpha(\Delta)$	the concatenation of all $n_{\text{add}}$ words $\alpha^1, \ldots, \alpha^{n_{\text{add}}}$ ex-
	cluding their MSBs
$oldsymbol{eta}(\Delta)$	the concatenation of all $n_{\text{add}}$ words $\beta^1, \ldots, \beta^{n_{\text{add}}}$ ex-
	cluding their MSBs
$egin{aligned} oldsymbol{lpha}_k, oldsymbol{eta}_k, oldsymb$	the k-th bit of the vectors $\alpha$ , $\beta$ , $A$ , $B$ , $C$
$g^{\kappa}$	the k-th NUBFs of a binary-FSM-based Compress function
$g_{ m lin}^{\kappa}$	the linear approximation of $g^k$ as the one in Compress <sub>lin</sub>
	for a binary-FSM-based Compress function
$x^k$	the input vector of the k-th NUBFs of a binary-FSM-
	based compression function (i.e. $g^k$ ) which computes
	Compress(M,V)
$\delta^k$	the input vector of the $k$ -th linearized NUBFs of a linearized
	binary-FSM-based compression function (i.e. $g_{lin}^k$ ) which
	computes $Compress_{lin}(\Delta)$
$\Lambda(M,V)$	the output of all the $n_{nl}$ NUBFs of a binary-FSM-based
\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \	Compress function
$\Phi(\Delta)$	the output of all the $n_{\rm nl}$ linearized NUBFs of a linearized
` ′	binary-FSM-based compression function Compress <sub>lin</sub>
$\Lambda^{\Delta}(M,V)$	see Section 5
$\Gamma(\Delta)$	see Section 5
$\Lambda_k(M,V)$	the k-th bit of the vector $\Lambda(M, V)$ which equals $g^k(x^k)$
	the k-th bit of the vector $\Phi(\Delta)$ which equals $g_{\text{lin}}^k(\delta^k)$
$\frac{\Phi_k(\Delta)}{\Lambda_k^{\Delta}(M,V)}$	the $h$ -th bit of the vector $\Phi(\Delta)$ which equals $g_{\text{lin}}(0)$
$A_k(M,V)$	the k-th bit of the vector $\Lambda^{\Delta}(M, V)$ which equals
	$g^k(x^k \oplus \delta^k)$
$\Gamma_k(\Delta)$	the k-th bit of the vector $\Gamma(\Delta)$ ; $\Gamma_k(\Delta) = 1$ iff $\delta^k \neq 0$

It is not difficult to prove that under some reasonable independence assumptions  $p_{\Delta}$ , which we call conforming probability, is the probability that a random message M follows the trail of  $\Delta$ . This is a direct corollary of Lemma 1 and Definition 1 . The exact proof can be done by induction on  $n_{\rm add}$ , the number of additions in the compression function. Due to other possible non-conforming pairs that start from message difference  $\Delta$  and lead to output difference Compress $_{\rm lin}(\Delta)$ ,  $p_{\Delta}$  is a lower bound for the desired probability in the lemma.

If  $\mathsf{Compress_{lin}}(\Delta)$  is of low Hamming weight, we get a near collision in the output. The interesting  $\Delta$ 's for collision search are those which belong to the kernel of  $\mathsf{Compress_{lin}}$ , *i.e.* those that satisfy  $\mathsf{Compress_{lin}}(\Delta) = 0$ . From now on, we assume that  $\Delta \neq 0$  is in the kernel of  $\mathsf{Compress_{lin}}$ , hence looking for collisions. According to Lemma 2, one needs to try around  $1/p_\Delta$  random message pairs in order to find a collision which conforms to the trail of  $\Delta$ . However in a random search it is better not to restrict oneself to the conforming messages as a collision at the end is all we want. Since  $p_\Delta$  is a lower bound for the probability of getting a collision for a message pair with difference  $\Delta$ , we might get a collision sooner. In Section 3 we explain a method which might find a conforming message by avoiding random search.

#### 2.2 Link with coding theory

We would like to conclude this section with a note on the relation between the following two problems:

- I) finding low-weight codewords of a linear code,
- II) finding a high probability linear differential path.

Since the functions  $\mathsf{Compress}_{\mathsf{lin}}(\Delta)$ ,  $\alpha(\Delta)$  and  $\beta(\Delta)$  are linear, we consider  $\Delta$  as a column vector and attribute three matrices  $\mathcal{H}$ ,  $\mathcal{A}$  and  $\mathcal{B}$  to these three transformations, respectively. In other words we have  $\mathsf{Compress}_{\mathsf{lin}}(\Delta) = \mathcal{H}\Delta$ ,  $\alpha(\Delta) = \mathcal{A}\Delta$  and  $\beta(\Delta) = \mathcal{B}\Delta$ . We then call  $\mathcal{H}$  the *parity check* matrix of the compression function.

Based on an initial work by Chabaud and Joux [11], the link between these two problems has been discussed by Rijmen and Oswald in [26] and by Pramstaller et al. in [25] with the general conclusion that finding highly probable differential paths is related to low weight codewords of the attributed linear code. In fact the relation between these two problems is more delicate. For problem (I), we are provided with the parity check matrix  $\mathcal{H}$  of a linear code for which a codeword  $\Delta$  satisfies the relation  $\mathcal{H}\Delta=0$ . Then, we are supposed to find a low-weight nonzero codeword  $\Delta$ . This problem is believed to be hard and there are some heuristic approaches for it, see [10] for example. For problem (II), however, we are given three matrices  $\mathcal{H}$ ,  $\mathcal{A}$  and  $\mathcal{B}$  and need to find a nonzero  $\Delta$  such that  $\mathcal{H}\Delta=0$  and  $\mathcal{A}\Delta\vee\mathcal{B}\Delta$  is of low-weight, see Lemma 2. Nevertheless, low-weight codewords  $\Delta$ 's matrix  $\mathcal{H}$  might be good candidates for providing low-weight  $\mathcal{A}\Delta\vee\mathcal{B}\Delta$ , i.e. differential paths with high probability  $p_\Delta$ . In particular, this approach is promising if these three matrices are sparse.

# 3 Finding a conforming message pair efficiently

The methods that are used to accelerate the finding of a message which satisfies some requirements are referred to as *freedom degrees use* in the literature. This includes message modifications [27], neutral bits [7], boomerang attacks [16, 20], tunnels [18] and submarine modifications [21]. In this section we show that the problem of finding conforming message pairs can be reformulated as finding preimages of zero under a function which we call the *condition* function. One can carefully analyze the condition function to see how freedom degrees might be used in efficient preimage reconstruction. Our method is based on measuring the amount of influence which every input bit has on each output bit of the condition function. We introduce the dependency tables to distinguish the influential bits, from those which have no influence or are less influential. In other words, in case the condition function does not mix its input bits well, we profit not only from neutral bits [7] but also from probabilistic neutral bits [2]. This is achieved by devising a backtracking search algorithm, similar to [6, 14, 24], based on the dependency table.

#### 3.1 Condition function

Let's assume that we have a differential path for the message difference  $\Delta$  which holds with probability  $p_\Delta = 2^{-y}$ . According to Lemma 2 we have  $y = \operatorname{wt} \left( \alpha(\Delta) \vee \beta(\Delta) \right)$ . In this section we show that, given an initial value V, the problem of finding a conforming message pair such that  $\operatorname{Compress}(M,V) \oplus \operatorname{Compress}(M \oplus \Delta,V) = 0$  can be translated into finding a message M such that  $\operatorname{Condition}_\Delta(M,V) = 0$ . Here  $Y = \operatorname{Condition}_\Delta(M,V)$  is a function which maps m-bit message M and v-bit initial value V into y-bit output Y. In other words, the problem is reduced to finding a preimage of zero under the  $\operatorname{Condition}_\Delta$  function. As we will see it is quite probable that not every output bit of the  $\operatorname{Condition}_\Delta$  function depends on all the message input bits. By taking a good strategy, this property enables us to find the preimages under this function more efficiently than random search. But of course, we are only interested in preimages of zero. In order to explain how we derive the function  $\operatorname{Condition}_T$  for  $\operatorname{Compress}_T$  we first present a quite easy-to-prove lemma. We recall that the  $\operatorname{carry}_T$  word of two words A and B is defined as  $C = (A + B) \oplus A \oplus B$ .

**Lemma 3.** Let A and B be two n-bit words and C represent their carry word. Let  $\delta = 2^i$  for  $0 \le i \le n-2$ . Then,

$$((A \oplus \delta) + (B \oplus \delta)) = (A + B) \Leftrightarrow A_i \oplus B_i \oplus 1 = 0,$$
 (2)

$$(A + (B \oplus \delta)) = (A + B) \oplus \delta \Leftrightarrow A_i \oplus C_i = 0,$$
(3)

and similarly

$$((A \oplus \delta) + B) = (A + B) \oplus \delta \Leftrightarrow B_i \oplus C_i = 0.$$
 (4)

For a given difference  $\Delta$ , a message M and an initial value V, let  $\mathbf{A}_k$ ,  $\mathbf{B}_k$ ,  $\mathbf{C}_k$ ,  $\alpha_k$  and  $\beta_k$ ,  $0 \le k < (n-1)n_{\mathrm{add}}$ , respectively denote the k-th bit of the output vectors of the functions  $\mathbf{A}(M,V)$ ,  $\mathbf{B}(M,V)$ ,  $\mathbf{C}(M,V)$ ,  $\alpha(\Delta)$  and  $\beta(\Delta)$ , as defined

in Section 2.1. Let  $\{i_0,\ldots,i_{y-1}\}$ ,  $0 \le i_0 < i_1 < \cdots < i_{y-1} < (n-1)n_{\mathrm{add}}$  be the positions of 1's in the vector  $\boldsymbol{\alpha} \vee \boldsymbol{\beta}$ . We define the function  $Y = \mathsf{Condition}_{\Delta}(M,V)$  as:

$$Y_{j} = \begin{cases} \mathbf{A}_{i_{j}} \oplus \mathbf{B}_{i_{j}} \oplus 1 & \text{if } (\boldsymbol{\alpha}_{i_{j}}, \boldsymbol{\beta}_{i_{j}}) = (1, 1), \\ \mathbf{A}_{i_{j}} \oplus \mathbf{C}_{i_{j}} & \text{if } (\boldsymbol{\alpha}_{i_{j}}, \boldsymbol{\beta}_{i_{j}}) = (0, 1), \\ \mathbf{B}_{i_{j}} \oplus \mathbf{C}_{i_{j}} & \text{if } (\boldsymbol{\alpha}_{i_{j}}, \boldsymbol{\beta}_{i_{j}}) = (1, 0), \end{cases}$$

$$(5)$$

for  $j = 0, 1, \dots, y - 1$ . This equation can be equivalently written as equation (8).

**Proposition 1.** For a given V and  $\Delta$ , a message M conforms to the trail of  $\Delta$  iff  $Condition_{\Delta}(M,V)=0$ .

*Proof.* The proof is straightforward from Definition 1, Lemma 3 and the definition of Condition function in equation (5).

### 3.2 Dependency table for freedom degrees use

For simplicity and generality, let's adopt the notation  $F(M, V) = \text{Condition}_{\Delta}(M, V)$ in this section. Assume that we are given a general function Y = F(M, V) which maps m message bits and v initial value bits into y output bits. Our goal is to reconstruct preimages of a particular output, for example the zero vector, efficiently. More precisely, we want to find V and M such that F(M, V) = 0. If F mixes its input bits very well, one needs to try about  $2^y$  random inputs in order to find one mapping to zero. However, in some special cases, not every input bit of F affects every output bit. Consider an ideal situation where message bits and output bits can be divided into  $\ell$  and  $\ell+1$  disjoint subsets respectively as  $\bigcup_{i=1}^{\ell} \mathcal{M}_i$  and  $\bigcup_{i=0}^{\ell} \mathcal{Y}_i$  such that the output bits  $\mathcal{Y}_j$   $(0 \le j \le \ell)$ only depend on the input bits  $\bigcup_{i=1}^{j} \mathcal{M}_{i}$  and the initial value V. In other words, once we know the initial value V, we can determine the output part  $\mathcal{Y}_0$ . If we know the initial value V and the input portion  $\mathcal{M}_1$ , the output part  $\mathcal{Y}_1$  is then known and so on. Refer to Section 6 to see the partitioning of a condition function related to MD6. This property of F suggests Algorithm 1 for finding a preimage of zero. Algorithm 1 is a backtracking search algorithm in essence, similar to [6, 14, 24], and in practice is implemented recursively with a tree-based search to avoid memory requirements. The values  $q_0, q_1, \dots, q_\ell$  are the parameters of the algorithm to be determined later. To discuss the complexity of the algorithm, let  $|\mathcal{M}_i|$  and  $|\mathcal{Y}_i|$  denote the cardinality of  $\mathcal{M}_i$ and  $\mathcal{Y}_i$  respectively, where  $|\mathcal{Y}_0| \geq 0$  and  $|\mathcal{Y}_i| \geq 1$  for  $1 \leq i \leq \ell$ . We consider an *ideal* behavior of F for which each output part depends in a complex way on all the variables that it depends on. Thus, the output segment changes independently and uniformly at random if we change any part of the relevant input bits.

To analyze the algorithm, we need to compute the optimal values for  $q_0,\ldots,q_\ell$ . The time complexity of the algorithm is  $\sum_{i=0}^\ell 2^{q_i}$  as at each step  $2^{q_i}$  values are examined. The algorithm is successful if we have at least one candidate left at the end, i.e.  $q'_{\ell+1} \geq 0$ . We have  $q'_{i+1} \approx q_i - |\mathcal{Y}_i|$ , coming from the fact that at the i-th step  $2^{q_i}$  values are examined each of which makes the portion  $\mathcal{Y}_i$  of the output null with probability  $2^{-|\mathcal{Y}_i|}$ . Note that we have the restrictions  $q_i - q'_i \leq |\mathcal{M}_i|$  and  $0 \leq q'_i$  since we have  $|\mathcal{M}_i|$  bits

# Algorithm 1: Preimage finding

```
Require: q_0, q_1, \ldots, q_\ell
Ensure: some preimage of zero under F

0: Choose 2^{q_0} initial values at random and keep those 2^{q_1'} candidates which make \mathcal{Y}_0 part null.

1: For each candidate, choose 2^{q_1-q_1'} values for \mathcal{M}_1 and keep those 2^{q_2'} ones making \mathcal{Y}_1 null.

2: For each candidate, choose 2^{q_2-q_2'} values for \mathcal{M}_2 and keep those 2^{q_3'} ones making \mathcal{Y}_2 null.

i:

i: For each candidate, choose 2^{q_i-q_\ell'} values for \mathcal{M}_i and keep those 2^{q_i'+1} ones making \mathcal{Y}_i null.

i:

\ell: For each candidate, choose 2^{q_\ell-q_\ell'} values for \mathcal{M}_\ell and keep those 2^{q_\ell'+1} final candidates making \mathcal{Y}_\ell null.
```

of freedom degree at the *i*-th step and we require at least one surviving candidate after each step. Hence, the optimal values for  $q_i$ 's can be recursively computed as  $q_{i-1} = |\mathcal{Y}_{i-1}| + \max(0, q_i - |\mathcal{M}_i|)$  for  $i = \ell, \ell - 1, \ldots, 1$  with  $q_\ell = |\mathcal{Y}_\ell|$ .

How can we determine the partitions  $\mathcal{M}_i$  and  $\mathcal{Y}_i$  for a given function F? We propose the following heuristic method for determining the message and output partitions in practice. We first construct a  $y \times m$  binary valued table T called *dependency table*. The entry  $T_{i,j}$ ,  $0 \le i \le m-1$  and  $0 \le j \le y-1$ , is set to one iff the j-th output bit is highly affected by the i-th message bit. To this end we empirically measure the probability that changing the i-th message bit changes the j-th output bit. The probability is computed over random initial values and messages. We then set  $T_{i,j}$  to one iff this probability is greater than a threshold  $0 \le th < 0.5$ , for example th = 0.3. We then call Algorithm 2.

# Algorithm 2: Message and output partitioning

```
Require: Dependency table T
Ensure: \ell, message partitions \mathcal{M}_1, \ldots, \mathcal{M}_{\ell} and output partitions \mathcal{Y}_0, \ldots, \mathcal{Y}_{\ell}.
 1: Put all the output bits j in \mathcal{Y}_0 for which the row j of T is all-zero.
 2: Delete all the all-zero rows from T.
 3: \ell := 0;
 4: while T is not empty do
 5:
        \ell := \ell + 1;
 6:
        repeat
 7:
           Determine the column i in T which has the highest number of 1's and delete it from T.
 8:
           Put the message bit which corresponds to the deleted column i into the set \mathcal{M}_{\ell}.
 9:
        until There is at least one all-zero row in T OR T becomes empty
        If T is empty set \mathcal{Y}_{\ell} to those output bits which are not in \bigcup_{i=0}^{\ell-1} \mathcal{Y}_i and stop.
10:
        Put all the output bits j in \mathcal{Y}_{\ell} for which the corresponding row of T is all-zero.
11:
12:
        Delete all the all-zero rows from T.
13: end while
```

In practice, once we make a partitioning for a given function using the above method, there are two issues which may cause the ideal behavior assumption to be violated:

- 1. The message segments  $\mathcal{M}_1, \dots, \mathcal{M}_i$  do not have full influence on  $\mathcal{Y}_i$ ,
- 2. The message segments  $\mathcal{M}_{i+1}, \ldots, \mathcal{M}_{\ell}$  have influence on  $\mathcal{Y}_0, \ldots, \mathcal{Y}_i$ .

With regard to the first issue, we ideally would like that all the message segments  $\mathcal{M}_1, \mathcal{M}_2, \dots, \mathcal{M}_i$  as well as the initial value V have full influence on the output part  $\mathcal{Y}_i$ . In practice the effect of the last few message segments  $\mathcal{M}_{i-d_i}, \dots, \mathcal{M}_i$  (for some small integer  $d_i$ ) is more important, though. Theoretical analysis of deviation from this requirement may not be easy. However, with some tweaks on the tree-based (backtracking) search algorithm, we may overcome this effect in practice. For example if the message segment  $\mathcal{M}_{i-1}$  does not have a great influence on the output segment  $\mathcal{Y}_i$ , we may decide to backtrack two steps at depth i, instead of one (the default value). The reason is as follows. Imagine that you are at depth i of the tree and you are trying to adjust the *i*-th message segment  $\mathcal{M}_i$ , to make the output segment  $\mathcal{Y}_i$  null. If after trying about  $2^{\min(|\mathcal{M}_i|,|\mathcal{Y}_i|)}$  choices for the *i*-th message block, you do not find an appropriate one, you will go one step backward and choose another choice for the (i-1)-st message segment  $\mathcal{M}_{i-1}$ ; you will then go one step forward once you have successfully adjusted the (i-1)-st message segment. If  $\mathcal{M}_{i-1}$  has no effect on  $\mathcal{Y}_i$ , this would be useless and increase our search cost at this node. Hence it would be appropriate if we backtrack two steps at this depth. In general, we may tweak our tree-based search by setting the number of steps which we want to backtrack at each depth.

In contrast, the theoretical analysis of the second issue is easy. Ideally, we would like that the message segments  $\mathcal{M}_i, \ldots, \mathcal{M}_\ell$  have no influence on the output segments  $\mathcal{Y}_0, \dots, \mathcal{Y}_{i-1}$ . The smaller the threshold value th is chosen, the less the influence would be. Let  $2^{-p_i}$ ,  $1 \le i \le \ell$ , denote the probability that changing the message segment  $\mathcal{M}_i$  does not change any bit from the output segments  $\mathcal{Y}_0, \dots, \mathcal{Y}_{i-1}$ . The probability is computed over random initial values and messages, and a random non-zero difference in the message segment  $\mathcal{M}_i$ . Algorithm 1 must be reanalyzed in order to recompute the optimal values for  $q_0, \ldots, q_\ell$ . Algorithm 1 also needs to be slightly changed by reassuring that at step i, all the output segments  $\mathcal{Y}_0, \dots, \mathcal{Y}_{i-1}$  remain null. The time complexity of the algorithm is still  $\sum_{i=0}^{\ell} 2^{q_i}$  and it is successful if at least one surviving candidate is left at the end, i.e.  $q_{\ell+1} \geq 0$ . However, here we set  $q'_{i+1} \approx q_i - |\mathcal{Y}_i| - p_i$ . This comes from the fact that at the *i*-th step  $2^{q_i}$  values are examined each of which makes the portion  $\mathcal{Y}_i$  of the output null with probability  $2^{-|\mathcal{Y}_i|}$ and keeping the previously set output segments  $\mathcal{Y}_0, \ldots, \mathcal{Y}_{i-1}$  null with probability  $2^{-p_i}$ (we assume these two events are independent). Here, our restrictions are again  $0 \le q_i'$ and  $q_i - q_i' \leq |\mathcal{M}_i|$ . Hence, the optimal values for  $q_i$ 's can be recursively computed as  $q_{i-1} = p_{i-1} + |\mathcal{Y}_{i-1}| + \max(0, q_i - |\mathcal{M}_i|)$  for  $i = \ell, \ell - 1, \dots, 1$  with  $q_\ell = |\mathcal{Y}_\ell|$ .

Remark 1. When working with functions with a huge number of input bits, it might be appropriate to consider the m-bit message M as a string of u-bit units instead of bits. For example one can take u=8 and work with bytes. We then use the notation  $M=(M[0],\ldots,M[m/u-1])$  (assuming u divides m) where  $M[i]=(M_{iu},\ldots,M_{iu+u-1})$ . In this case the dependency table must be constructed according to the probability that changing every message unit changes each output bit.

# 4 Application to CubeHash

CubeHash [4] is Bernstein's proposal for the NIST SHA-3 competition [22]. CubeHash variants, denoted by CubeHash-r/b, are parametrized by r and b which at each iteration process b bytes in r rounds. Although CubeHash-8/1 was the original official submission, later the designer proposed the tweak CubeHash-16/32 which is almost 16 times faster than the initial proposal [5]. Nevertheless, the author has encouraged cryptanalysis of CubeHash-r/b variants for smaller r's and bigger b's.

### 4.1 CubeHash description

CubeHash works with 32-bit words (n=32) and uses three simple operations: XOR, rotation and modular addition. It has an internal state  $S=(S_0,S_1,\ldots,S_{31})$  of 32 words and its variants, denoted by CubeHash-r/b, are identified by two parameters  $r\in\{1,2,\ldots\}$  and  $b\in\{1,2,\ldots,128\}$ . The internal state S is set to a specified value which depends on the digest length (limited to 512 bits) and parameters r and b. The message to be hashed is appropriately padded and divided into b-byte message blocks. At each iteration one message block is processed as follows. The 32-word internal state S is considered as a 128-byte value and the message block is XORed into the first b bytes of the internal state<sup>5</sup>. Then, the following fixed permutation is applied r times to the internal state to prepare it for the next iteration.

- 1. Add  $S_i$  into  $S_{i\oplus 16}$ , for  $0 \le i \le 15$ .
- 2. Rotate  $S_i$  to the left by seven bits, for  $0 \le i \le 15$ .
- 3. Swap  $S_i$  and  $S_{i \oplus 8}$ , for  $0 \le i \le 7$ .
- 4. XOR  $S_{i\oplus 16}$  into  $S_i$ , for  $0 \le i \le 15$ .
- 5. Swap  $S_i$  and  $S_{i\oplus 2}$ , for  $i \in \{16, 17, 20, 21, 24, 25, 28, 29\}$ .
- 6. Add  $S_i$  into  $S_{i\oplus 16}$ , for  $0 \le i \le 15$ .
- 7. Rotate  $S_i$  to the left by eleven bits, for  $0 \le i \le 15$ .
- 8. Swap  $S_i$  and  $S_{i\oplus 4}$ , for  $i \in \{0, 1, 2, 3, 8, 9, 10, 11\}$ .
- 9. XOR  $S_{i\oplus 16}$  into  $S_i$ , for  $0 \le i \le 15$ .
- 10. Swap  $S_i$  and  $S_{i\oplus 1}$ , for  $i \in \{16, 18, 20, 22, 24, 26, 28, 30\}$ .

Having processed all message blocks, a fixed transformation is applied to the final internal state to extract the hash value as follows. First, the last state word  $S_{31}$  is ORed with integer 1 and then the above permutation is applied  $10 \times r$  times to the resulting internal state. Finally, the internal state is truncated to produce the message digest of desired hash length. Refer to [4] for the full specification.

<sup>&</sup>lt;sup>5</sup> The first message byte into the least significant byte of  $S_0$ , the second one into the second least significant byte of  $S_0$ , the third one into the third least significant byte of  $S_0$ , the fourth one into the most significant byte of  $S_0$ , the fifth one into the least significant byte of  $S_1$ , and so forth until all b message bytes have been exhausted.

### 4.2 Definition of the compression function Compress

To be in the line of our general method, we need to deal with fixed-size input compression functions. To this end, we consider t ( $t \geq 1$ ) consecutive iterations of Cube-Hash. We define the function  $H = \mathsf{Compress}(M,V)$  with an 8bt-bit message  $M = M^0 || \dots || M^{t-1}$ , a 1024-bit initial value V and a (1024-8b)-bit output H. The initial value V is used to initialize the 32-word internal state of CubeHash. Each  $M^i$  is a b-byte message block. We start from the initialized internal state and update it in t iterations. That is, in t iterations the t message blocks  $M^0, \dots, M^{t-1}$  are sequentially processed in order to transform the internal state into a final value. The output H is then the last 128-b bytes of the final internal state value which is ready to absorb the (t+1)-st message block (the 32-word internal state is interpreted as a 128-byte vector).

Our goal is to find collisions for this Compress function. In the next section we explain how collisions can be constructed for CubeHash itself.

#### 4.3 Collision construction

Once we find message prefix  $M^{\mathrm{pre}}$ , message M and difference  $\Delta$ , any message pairs (M',M'') of the above-mentioned form is a collision for CubeHash for any message block  $M^t$  and any message suffix  $M^{\mathrm{suf}}$ . We find the difference  $\Delta$  using the linearization method of Section 2 to applied to CubeHash in the next section. Then,  $M^{\mathrm{pre}}$  and M are found by finding a preimage of zero under the Condition function as explained in Section 3. Algorithm 4 in Appendix F shows how CubeHash Condition function can be implemented in practice for a given differential path.

### 4.4 Linear differentials for CubeHash-r/b

As we explained in Section 2, the linear transformation Compress<sub>lin</sub> can be identified by a matrix  $\mathcal{H}_{h\times m}$ . We are interested in  $\Delta$ 's such that  $\mathcal{H}\Delta=0$  and such that the differential trails have high probability. For CubeHash-r/b with t iterations,  $\Delta=\Delta^0||\dots||\Delta^{t-1}$  and  $\mathcal{H}$  has size  $(1024-8b)\times 8bt$ , see Section 4.2. This matrix suffers from having low rank. This enables us to find low weight vectors of the kernel. We then hope that they are also good candidates for providing highly probable trails, see

Section 2.2. Assume that this matrix has rank  $(8bt - \tau)$ ,  $\tau \geq 0$ , signifying existence of  $2^{\tau} - 1$  nonzero solutions to  $\mathcal{H}\Delta = 0$ . To find a low weight nonzero  $\Delta$ , we use the following method.

The rank of  $\mathcal{H}$  being  $(8bt - \tau)$  shows that the solutions can be expressed by identifying  $\tau$  variables as free and expressing the rest in terms of them. Any choice for the free variables uniquely determines the remaining  $8bt - \tau$  variables, hence providing a unique member of the kernel. We choose a set of  $\tau$  free variables at random. Then, we set one, two, or three of the  $\tau$  free variables to bit value 1, and the other  $\tau - 1$ , or  $\tau - 2$  or  $\tau-3$  variables to bit value 0 with the hope to get a  $\Delta$  providing a high probability differential path. We have made exhaustive search over all  $\tau + {\tau \choose 2} + {\tau \choose 3}$  possible choices for all  $b \in \{1, 2, 3, 4, 8, 16, 32, 48, 64\}$  and  $r \in \{1, 2, 3, 4, 5, 6, 7, 8\}$  in order to find the best characteristics. Table 2 includes the ordered pair (t, y), i.e. the corresponding number of iterations and the  $-\log_2$  probability (number of bit conditions) of the best raw probability path we found. For most of the cases, the best characteristic belongs to the minimum value of t for which  $\tau > 0$ . There are a few exceptions to consider which are starred in Table 2. For example in the CubeHash-3/4 case, while for t=2 we have  $\tau=4$  and y=675, by increasing the number of iterations to t=4, we get  $\tau=40$ and a better characteristic with y = 478. This may hold for other cases as well since we only increased t until our program terminated in a reasonable time. We would like to emphasize that since we are using linear differentials, the erasing block difference  $\Delta^t$ only depends on the difference  $\Delta$ , see Section 4.3.

**Table 2.** The values of (t, y) for the differential path with the best found raw probability.

$r \setminus b$	1	2	3	4	8	12	16	32	48	64
1	(14, 1225)	$(8,221)^*$	(4, 46)	(4, 32)	(4, 32)	-	-	-	-	-
2	(7, 1225)	$(4,221)^*$	(2, 46)	(2, 32)	(2, 32)	-	-	_	_	-
3	$(16, 4238)^*$	(6, 1881)	(4,798)	$(4,478)^*$	$(4,478)^*$	$(4,400)^*$	$(4,400)^*$	$(4,400)^*$	$(3,364)^*$	(2,65)
4	(8, 2614)	(3,964)	(2, 195)	(2, 189)	(2, 189)	(2, 156)	(2, 156)	(2, 156)	(2, 130)	(2, 130)
5	$(18, 10221)^*$	(8,4579)	(4, 2433)	(4, 1517)	(4, 1517)	(4, 1244)	(4, 1244)	(4, 1244)	$(4, 1244)^*$	(2,205)
6	(10, 4238)	(3, 1881)	(2,798)	(2,478)	(2,478)	(2,400)	(2,400)	(2,400)	(2,351)	(2,351)
7	(14, 13365)	(8,5820)	(4,3028)	(4,2124)	(4,2124)	(4, 1748)	(4, 1748)	(4, 1748)	$(4, 1748)^*$	(2,447)
8	(4, 2614)	(4, 2614)	(2, 1022)	(2, 1009)	(2, 1009)	(2,830)	(2,830)	(2,830)	(2,637)	(2,637)

Second preimage attacks on CubeHash. Any differential path with raw probability greater than  $2^{-512}$  can be considered as a (theoretical) second preimage attack on CubeHash with 512-bit digest size. In Table 2 the entries which do not correspond to a successful second preimage attack, *i.e.* y > 512, are shown in gray, whereas the others have been highlighted. For example, our differential path for CubeHash-6/4 with raw probability  $2^{-478}$  indicates that by only one hash evaluation we can produce a second preimage with probability  $2^{-478}$ . Alternatively, it can be stated that for a fraction of  $2^{-478}$  messages we can easily provide a second preimage. The list of differential trails for highlighted entries can be found in Appendix A.

#### 4.5 Collision attacks on CubeHash variants

Although Table 2 includes our best found differential paths with respect to raw probability or equivalently second preimage attack, when it comes to freedom degrees use for collision attack, these trails might not be the optimal ones. In other words, for a specific r and b, there might be another differential path which is worse in terms of raw probability but is better regarding the collision attack complexity if we use some freedom degrees speedup. As an example, for CubeHash-3/48 with the path which has raw probability  $2^{-364}$ , using our method of Section 3 the time complexity can be reduced to about  $2^{58.9}$  (partial) evaluation of its condition function. However, there is another path with raw probability  $2^{-368}$  which has time complexity of about  $2^{53.3}$  (partial) evaluation of its condition function. Table 3 shows the best paths we found regarding the reduced complexity of the collision attack using our method of Section 3. While most of the paths are still the optimal ones with respect to the raw probability, the starred entries indicate the ones which invalidate this property. Some of the interesting differential paths for starred entries in Table 3 are given in Appendix B.

Table 4 shows the reduced time complexities of collision attack using our method of Section 3 for the differential paths of Table 3. To construct the dependency table, we have analyzed the Condition function at byte level, see Remark 1. The time complexities are in logarithm 2 basis and might be improved if the dependency table is analyzed at a bit level instead. The complexity unit is (partial) evaluation of their respective Condition function. We remind that the full evaluation of a Condition function corresponding to a t-iteration differential path is almost the same as application of t iterations (rt rounds) of CubeHash. We emphasize that the complexities are independent of digest size. All the complexities which are less than t0 bits. The complexities bigger than t0 bits. The complexities bigger than t0 bits digest size. The successfully attacked instances have been highlighted.

**Table 3.** The values of (t, y) for the differential path with the best found total complexity (Table 4 includes the reduced complexities using our method of Section 3).

\ 7.	1	0	2	4	0	10	1.0	20	40	C 4
$r \setminus b$	1		3	4	8	12	16	32	48	64
1	(14, 1225)	(8, 221)	(4, 46)	(4, 32)	(4, 32)	-	-	-	-	-
2	(7, 1225)	(4, 221)	(2,46)	(2, 32)	(2, 32)	-	-	-	-	-
3	(16, 4238)	(6, 1881)	(4,798)	(4,478)	(4,478)	(4,400)	(4,400)	(4,400)	$(3,368)^*$	(2,65)
4	(8, 2614)	(3,964)	(2, 195)	(2, 189)	(2, 189)	(2, 156)	(2, 156)	(2, 156)	$(2, 134)^*$	$(2, 134)^*$
5	(18, 10221)	(8,4579)	(4, 2433)	(4, 1517)	(4, 1517)	$(4, 1250)^*$	$(4, 1250)^*$	$(4, 1250)^*$	$(4, 1250)^*$	(2,205)
6	(10, 4238)	(3, 1881)	(2,798)	(2,478)	(2,478)	(2,400)	(2,400)	(2,400)	(2,351)	(2,351)
7	(14, 13365)							(4, 1748)		$(2,455)^*$
8	(4, 2614)	(4, 2614)	(2, 1022)	(2, 1009)	(2, 1009)	(2,830)	(2,830)	(2,830)	$(2,655)^*$	$(2,655)^*$

The astute reader should realize that the complexities of Table 4 correspond to the optimal threshold value, see Section 3.2. To see the effect of the threshold value th on the complexity, we focus on five instances of CubeHash: CubeHash-3/64,-4/48,-3/48,-3/32 and-4/32. The first two instances are the ones for which the theoretical complexi-

**Table 4.** Theoretical  $\log_2$  complexities of improved collision attacks with freedom degrees use at byte level for the differential paths of Table 3.

$r \setminus b$	1	2	3	4	8	12	16	32	48	64
	1121.0					-	-	-	-	-
	1177.0					-	-	-	-	-
3	4214.0	1793.0	720.0	380.1	292.6	153.5	102.0	55.6	53.3	9.4
	2598.0									
5	10085.0	4460.0	2345.0	1397.0	1286.0	946.0	868.0	588.2	425.0	71.7
	4230.0									
	13261.0									
8	2606.0	2590.0	982.0	953.0	889.0	699.0	662.0	524.3	313.0	304.4

ties are practically reachable and we have managed to find their real collision examples. The other three instances are the ones whose theoretical complexities are just above the practically reachable values and are most probably the ones for which real collisions will be found in near future either using more advanced methods or utilizing a huge cluster of computers. Table 5 shows the effect of threshold value on the complexity for these six instances.

**Table 5.** Theoretical  $\log_2$  complexities of improved attacks with freedom degrees use at byte level versus threshold value.

CubeHash instance	$y \setminus th$	0.0	0.025	0.05	0.1	0.15	0.2	0.25	0.3	0.35	0.4	0.45	0.475
3/64	65	16.3	11.3	11.4	10.6	10.2	10.4	9.9	9.4	10.4	10.4	11.4	16.5
4/48	134	50.8	38.3	34.6	33.6	32.3	38.4	35.0	32.2	30.7	32.3	37.7	38.6
4/32												63.9	
3/48	368	99.7	90.6	87.2	80.3	67.6	61.2	65.3	53.3	65.1	83.4	102.6	119.3
3/32	400	91.0	63.3	61.5	57.1	61.8	55.6	70.9	70.7	79.7	82.4	98.9	113.6

Real collisions for CubeHash-3/64 and -4/48. We use the 2-iteration message difference  $\Delta=\Delta^0||\Delta^1$  of relation (13) for CubeHash-3/64 and of relation (14) for CubeHash-4/48 (see Appendices A and B). These equations also include the erasing block difference  $\Delta^2$ , required for collision construction. Appendix C includes the values of  $M^{\rm pre}$  and  $M=M^0||M^1$  for collision on CubeHash-3/64 with 512-bit digest size, whereas Appendix D provides the corresponding values for collision constructing on CubeHash-4/48 with 512-bit digest size. In other words, Condition $_\Delta(M,V)=0$  holds for their corresponding condition functions where V is the content of the internal state after processing the message prefix  $M^{\rm pre}$ . According to Section 4.3, the pair M' and M'' where

$$\begin{split} M' &= M^{\rm pre} ||M^0||M^1||M^2||M^{\rm suf}, \\ M'' &= M^{\rm pre} ||(M^0 \oplus \varDelta^0)||(M^1 \oplus \varDelta^1)||(M^2 \oplus \varDelta^2)||M^{\rm suf}, \end{split}$$

collide for any message block  $M^2$  and any message suffix  $M^{\text{suf}}$  of arbitrary length. Note that a collision pair for a given r and b can be easily transformed to a collision pair for the same r and bigger b's by appending enough zeros to each message block.

**Practice versus theory.** We provided a framework which is handy in order to analyze many hash functions in a generic way. In practice, the optimal threshold value may be a little different from the theoretical one. Moreover, by slightly playing with the neighboring bits in the suggested partitioning corresponding to a given threshold value (Algorithm 2), we may achieve a partitioning which is more suitable for applying the attacks. In particular, Tables 4 and 5 contain the theoretical complexities for different CubeHash instances under the assumption that the Condition function behaves ideally with respect to the first issue discussed in Section 3.2. In practice, deviation from this assumption increases the effective complexity. For particular instances, more simulations need to be done to analyze the potential non-randomness effects in order to give a more exact estimation of the practical complexity.

In the following we compare the practical complexities with the theoretical values for some cases for which their complexities are practically reachable. Moreover, for some CubeHash instances for which their complexities are unreachable in practice we try to give a more precise estimation of their effective complexities.

Our tree-based search implementation for the <code>CubeHash-3/64</code> case with th=0.3 has median complexity  $2^{21}$  instead of the  $2^{9.4}$  of Table 5. The median decreases to  $2^{17}$  by backtracking three steps at each depth instead of one, see Section 3.2. We expect the practical complexities for other instances of CubeHash with three rounds to be slightly bigger than the theoretical numbers in Table 4. These cases need to be further investigated.

Our detailed analysis of <code>CubeHash-4/32</code>, <code>CubeHash-4/48</code> and <code>CubeHash-4/64</code> shows that these cases perfectly match with theory. According to Table 4, for <code>CubeHash-4/64</code> (with th=0.33) and <code>CubeHash-4/48</code> (with th=0.30) we have the theoretical complexities  $2^{28.8}$  and  $2^{30.7}$ , respectively. We experimentally achieve median complexities  $2^{28.3}$  and  $2^{30.4}$  respectively. For <code>CubeHash-4/32</code> (with th=0.25) the theoretical complexity is  $2^{54.7}$ . In the tree-based search algorithm, we need to satisfy 44 bit conditions at step 18, *i.e.*  $|\mathcal{Y}_{18}|=44$ . This is the node which has the highest cost and if it is successfully passed, the remaining steps will easily be followed. In other words, a single surviving candidate at this node (which means  $2^{q'_{19}}=1$  referring to Algorithm 1) suffices to make the remaining condition bits null with little cost. Our simulations show that on average we need about  $2^{10}$  (partial) evaluations of the condition function per one surviving candidate which arrives at depth 18. Hence, our estimation of the practical complexity is  $2^{10} \times 2^{44} = 2^{54}$  which agrees with theory.

In CubeHash-5/64 case (with th=0.24), the costliest node is at depth 20 for which 70 bit conditions must be satisfied, i.e.  $|\mathcal{Y}_{20}|=70$ . A single surviving candidate at this node suffices to make the remaining condition bits null with little cost. Our simulation shows that on average about  $2^{7.0}$  (partial) evaluations of the condition function are required per one surviving candidate which arrives at depth 20. Hence, our estimation of the practical complexity is about  $2^{7.0+70}=2^{77.0}$ , versus theoretical value  $2^{71.7}$ .

In CubeHash-6/16 case (with th=0.15), the costliest node is at depth 9 for which 198 bit conditions must be satisfied, i.e.  $|\mathcal{Y}_9|=198$ . We need  $2^{26}$  candidates to successfully pass this node, i.e.  $q'_{10}=26$ , see Algorithm 1. Our simulation shows that on average about  $2^5$  (partial) evaluations of the condition function are required per one surviving candidate which arrives at depth 9. Hence, our estimation of the practical complexity is about  $2^{26+5+198}=2^{229}$ , versus theoretical value  $2^{222.6}$ .

In CubeHash-7/64 case (with th=0.255), the costliest node is at depth 24 for which 201 bit conditions must be satisfied, i.e.  $|\mathcal{Y}_{24}|=201$ . Only one surviving candidate at this node suffices to make the remaining condition bits null with much less cost. Our simulation shows that on average about  $2^7$  (partial) evaluations of the condition function are required per one surviving candidate which arrives at depth 24. Hence, our estimation of the practical complexity is about  $2^{7+201}=2^{208}$ , versus theoretical value  $2^{203.0}$ .

We emphasize that for these latter cases we did not attempt to play with the neighboring bits in the partitioning. We believe, in general, complexities can get very close to the theoretical ones if one tries to do so.

Comparison with the previous results. The first analysis of CubeHash was proposed by Aumasson et al. [3] in which the authors showed some non-random properties for several versions of CubeHash. A series of collision attacks on CubeHash-1/b and CubeHash-2/b for large values of b were announced by Aumasson [1] and Dai [12]. Collision attacks were later investigated deeply by Brier and Peyrin [8]. Our results improve on all existing ones as well as attacking some untouched variants.

#### 5 Generalization

In sections 2 and 3 we considered modular-addition-based compression functions which use only modular additions and linear transformations. Moreover, we concentrated on XOR approximation of modular additions in order to linearize the compression function. This method is however quite general and can be applied to a broad class of hash constructions, covering many of the existing hash functions. Additionally, it lets us consider other linear approximations as well. We view a compression function  $H = \operatorname{Compress}(M,V): \{0,1\}^m \times \{0,1\}^v \to \{0,1\}^h$  as a binary finite state machine (FSM). The FSM has an internal state which is consecutively updated using message M and initial value V. We assume that FSM operates as follows, and we refer to such Compress functions as binary-FSM-based. The concept can also cover non-binary fields.

The internal state is initially set to zero. Afterwards, the internal state is sequentially updated in a limited number of steps. The output value H is then derived by truncating the final value of the internal state to the specified output size. At each step, the internal state is updated according to one of these two possibilities: either the whole internal state is updated as an affine transformation of the current internal state, M and V, or  $only\ one$  bit of the internal state is updated as a nonlinear Boolean function of the current internal state, M and V. Without loss of generality, we assume that all of the nonlinear updating Boolean functions (NUBF) have zero constant term (i.e. the output of zero vector is zero) and none of the involved variables appear as a pure linear

term (i.e. changing any input variable does not change the output bit with certainty). As we will see, this assumption, coming from the simple observation that we can integrate constants and linear terms in an affine updating transformation (AUT), is essential for our analysis. Linear approximations of the FSM can be achieved by replacing AUTs with linear transformations by ignoring the constant terms and NUBFs with linear functions of their arguments. Similar to Section 2 this gives us a linearized version of the compression function which we denote by  $\mathsf{Compress}_{\mathsf{lin}}(M,V)$ . As we are dealing with differential cryptanalysis, we take the notation  $\mathsf{Compress}_{\mathsf{lin}}(M) = \mathsf{Compress}_{\mathsf{lin}}(M,0)$ . The argument given in Section 2 is still valid: elements of the kernel of the linearized compression function (i.e.  $\Delta$ 's s.t.  $\mathsf{Compress}_{\mathsf{lin}}(\Delta) = 0$ ) can be used to construct differential trails.

Let  $n_{\rm nl}$  denote the total number of NUBFs in the FSM. We count the NUBFs by starting from zero. We introduce four functions  $\Lambda(M,V)$ ,  $\Phi(\Delta)$ ,  $\Lambda^{\Delta}(M,V)$  and  $\Gamma(\Delta)$ all of output size  $n_{\rm nl}$  bits. To define these functions, consider the two procedures which implement the FSMs of  $\mathsf{Compress}(M,V)$  and  $\mathsf{Compress}_{\mathrm{lin}}(\Delta)$ . Let the Boolean function  $g^k$ ,  $0 \le k < n_{\rm nl}$ , stand for the k-th NUBF and denote its linear approximation as in Compress $_{\mathrm{lin}}$  by  $g_{\mathrm{lin}}^k$ . Moreover, denote the input arguments of the Boolean functions  $g^k$  and  $g_{\mathrm{lin}}^k$  in the FSMs which compute  $\mathsf{Compress}(M,V)$  and  $\mathsf{Compress}_{\mathrm{lin}}(\Delta)$  by the vectors  $x^k$  and  $\delta^k$ , respectively. Note that  $\delta^k$  is a function of  $\Delta$  whereas  $x^k$  depends on M and V. The k-th bit of  $\Gamma(\Delta)$ ,  $\Gamma_k(\Delta)$ , is set to one iff the argument of the k-th linearized NUBF is not the all-zero vector, i.e.  $\Gamma_k(\Delta) = 1$  iff  $\delta^k \neq 0$ . The k-th bit of  $\Lambda(M,V)$ ,  $\Lambda_k(M,V)$ , is the output value of the k-th NUBF in the course of evaluation of Compress(M, V) through the execution of FSM, i.e.  $\Lambda_k(M, V) = g^k(x^k)$ . Similarly, the k-th bit of  $\Phi(\Delta)$ ,  $\Phi_k(\Delta)$ , is the output value of the k-th linearized NUBF in the course of evaluation of  $\mathsf{Compress}_{\mathsf{lin}}(\Delta)$  through linearized version of FSM, i.e.  $\Phi_k(\Delta) = g_{lin}^k(\delta^k)$ . The k-th bit of  $\Lambda^{\Delta}(M,V)$ ,  $\Lambda_k^{\Delta}(M,V)$ , is evaluated in a more complex way: apply the k-th nonlinear Boolean function on the XOR-difference of the arguments which are fed into the two Boolean functions which compute  $\Lambda_k(M,V)$ and  $\Phi_k(\Delta)$ , i.e.  $\Lambda_k^{\Delta}(M,V) = g^k(x^k \oplus \delta^k)$ . Refer to Table 1 for a list of these notations along with the ones used in sections 2 and 3.1. We can then present the following proposition.

**Proposition 2.** Let Compress be a binary-FSM-based compression function. For any message difference  $\Delta$ , let  $\{i_0,\ldots,i_{y-1}\}$ ,  $0 \le i_0 < i_1 < \cdots < i_{y-1} < n_{\rm nl}$  be the positions of 1's in the vector  $\Gamma(\Delta)$  where  $y = {\rm wt}(\Gamma(\Delta))$ . We define the condition function  $Y = {\rm Condition}_{\Delta}(M,V)$  where the j-th bit of Y is computed as

$$Y_j = \Lambda_{i_j}(M, V) \oplus \Lambda_{i_j}^{\Delta}(M, V) \oplus \Phi_{i_j}(\Delta). \tag{6}$$

Then, if  $\Delta$  is in the kernel of Compress<sub>lin</sub>, Condition<sub> $\Delta$ </sub>(M,V)=0 implies that the pair  $(M,M\oplus\Delta)$  is a collision for Compress with the initial value V.

*Proof.* First, we present give the following definition, a generalization of Definition 1.

**Definition 2.** We say that a message M (for a given V) conforms to (or follows) the trail of  $\Delta$  till before applying the i-th NUBF iff

$$\Lambda_k(M,V) \oplus \Lambda_k^{\Delta}(M,V) = \Phi_k(\Delta), \tag{7}$$

for  $0 \le k < i$ . Moreover, we say that the message M (for a given V) conforms to (or follows) the trail of  $\Delta$  iff the equation (7) holds for  $0 \le k < n_{\rm nl}$ .

To explain the Definition 2, imagine we have three procedures which implement the FSMs to compute  $\mathsf{Compress}(M,V)$ ,  $\mathsf{Compress}(M\oplus\Delta,V)$  and  $\mathsf{Compress}_{\mathsf{lin}}(\Delta)$ . If a message M (for a given V) conforms to the trial of  $\Delta$  till before applying the i-th NUBF, it means that at every step before the i-th NUBF the difference of the internal states of the two FSMs which compute  $\mathsf{Compress}(M,V)$  and  $\mathsf{Compress}(M\oplus\Delta,V)$  equals the internal state value of the FSM which computes  $\mathsf{Compress}_{\mathsf{lin}}(\Delta)$ . This is guaranteed by equation (7) because the only nonlinear operations during this period are the applications of the first i-1 NUBFs. Note that if a message M conforms to the trial of  $\Delta$  for a given V we then have  $\mathsf{Compress}(M,V) \oplus \mathsf{Compress}(M\oplus\Delta) = \mathsf{Compress}_{\mathsf{lin}}(\Delta)$ .

For simplicity, let the Boolean function g stand for the i-th NUBF and denote its linear approximation as in Compress $_{\rm lin}$  by  $g_{\rm lin}$ . Moreover, denote the input arguments of the Boolean functions g and  $g_{\rm lin}$  in the FSMs which compute Compress(M,V) and Compress $_{\rm lin}(\Delta)$  by the vectors x and  $\delta$ , respectively. According to our notations  $\Gamma_i(\Delta)=1$  iff  $\delta\neq 0$ , and  $\Lambda_i(M,V)=g(x)$ ,  $\Phi_i(\Delta)=g_{lin}(\delta)$  and  $\Lambda_i^\Delta(M,V)=g(x\oplus\delta)$ . Now, the role of  $\Gamma_i(\Delta)$  becomes visible.

If  $\Gamma_i(\Delta)=0$ , we have  $\delta=0$ , and hence  $g(x)\oplus g(x\oplus \delta)=g_{\mathrm{lin}}(\delta)$ . In this case the equation  $\Lambda_i(M,V)\oplus \Lambda_i^\Delta(M,V)\oplus \Phi_i(\Delta)=0$  is satisfied by itself, and therefore we do not need to introduce a condition bit, see equation (6). In other words the message M automatically conforms to the trial of  $\Delta$  for the initial value V till before applying the (i+1)-st NUBF, provided that it conforms to the trial of  $\Delta$  till before applying the i-th NUBF.

However, if  $\Gamma_i(\Delta)=1$  we have  $x\neq x\oplus \delta$ . Therefore, if the message M conforms to the trial of  $\Delta$  till before applying the i-th NUBF, it will also conform to the trial of  $\Delta$  till before applying the (i+1)-st NUBF iff  $g(x)\oplus g(x\oplus \delta)=g_{\mathrm{lin}}(\delta)$ . That is, we need to impose one condition bit as in equation (6). Remember we supposed that g has no linear term, the reason for which is as follows. If g has some linear terms and if the difference  $\delta$  is nonzero only in some of those linear inputs, the equation  $g(x)\oplus g(x\oplus \delta)=g_{\mathrm{lin}}(\delta)$  is held by itself. In this case, the imposed condition bit is redundant.

#### 5.1 Modular addition case

Let's review the compression functions involving only linear transformations and modular addition of n-bit words. We deeply studied this subject in sections 2 and 3. The modular addition Z=X+Y can be computed by considering one bit memory c for the carry bit. Let  $X=(x_0,\ldots,x_{n-1}), Y=(y_0,\ldots,y_{n-1})$  and  $Z=(z_0,\ldots,z_{n-1})$ . We have

$$c_{i+1} = c_i x_i \oplus c_i y_i \oplus x_i y_i$$
 for  $0 \le i \le n-2$   
 $z_i = x_i \oplus y_i \oplus c_i$  for  $0 \le i \le n-1$ ,

where  $c_0=0$ . It can be argued that a compression function which uses only linear transformations and  $n_{\rm add}$  modular additions can be implemented as a binary-FSM-based compression whose total number of NUBFs is  $n_{\rm nl}=(n-1)n_{\rm add}$ . All the NUBFs are of the form  $g(x,y,z)=xy\oplus xz\oplus yz$ . The XOR approximation of

modular addition in Section 2 corresponds to approximating all the NUBFs g by the zero function, i.e.  $g_{\text{lin}}(x,y,z)=0$ . Having the notations of sections 3.1, 2 and 5 in mind, see also Table 1, we deduce that the input argument of the k-th NUBF is  $(\mathbf{A}_k,\mathbf{B}_k,\mathbf{C}_k)$ ; whereas that of the k-th linearized NUBF is  $(\boldsymbol{\alpha}_k,\boldsymbol{\beta}_k,0)$ . Therefore we have  $A_k(M,V)=g(\mathbf{A}_k,\mathbf{B}_k,\mathbf{C}_k)$  and  $\Phi_k(\Delta)=g_{\text{lin}}(\boldsymbol{\alpha}_k,\boldsymbol{\beta}_k,0)$ . Moreover, we deduce  $\Gamma_k(\Delta)=\boldsymbol{\alpha}_k\vee\boldsymbol{\beta}_k\vee 0$  and  $A_k^\Delta(M,V)=g(\mathbf{A}_k\oplus\boldsymbol{\alpha}_k,\mathbf{B}_k\oplus\boldsymbol{\beta}_k,\mathbf{C}_k\oplus 0)$ . As a result we get

$$Y_{j} = \Lambda_{i_{j}}(M, V) \oplus \Lambda_{i_{j}}^{\Delta}(M, V) \oplus \Phi_{i_{j}}(\Delta)$$

$$= g(\mathbf{A}_{k}, \mathbf{B}_{k}, \mathbf{C}_{k}) \oplus g(\mathbf{A}_{k} \oplus \boldsymbol{\alpha}_{k}, \mathbf{B}_{k} \oplus \boldsymbol{\beta}_{k}, \mathbf{C}_{k} \oplus 0) \oplus 0$$

$$= (\boldsymbol{\alpha}_{i_{j}} \oplus \boldsymbol{\beta}_{i_{j}}) \mathbf{C}_{i_{j}} \oplus \boldsymbol{\alpha}_{i_{j}} B_{i_{j}} \oplus \boldsymbol{\beta}_{i_{j}} A_{i_{j}} \oplus \boldsymbol{\alpha}_{i_{j}} \boldsymbol{\beta}_{i_{j}}$$
(8)

whenever  $\alpha_{i_j} \vee \beta_{i_j} = 1$ ; this agrees with equation (5).

# 5.2 Note on the different linear approximations

Different combinations of different linear approximations of the NUBFs provide different linear approximations of the compression function. However, one should be careful to avoid approximations which might lead to contradictions due to dependency between different approximations. In fact the probability  $2^{-\Gamma(\Delta)}$  would not be a good estimate in this case if there are strong correlations between approximations. In the case of linear approximation of modular addition of n-bit words we have (n-1) NUBFs for the carry bits, out of which n-2 are of the form  $xy \oplus xc \oplus yc$  and one of the form xy (corresponding to the carry of the LSB). There are eight linear approximations for the earlier Boolean function (it effectively depends on three variables) and four linear approximations for the later one (it effectively depends on two variables). This shows the possibility of  $4 \times 8^{(n-2)}$  different linear approximations. For one particular linear approximation, if the difference of the two addends are  $\alpha$  and  $\beta$  the output difference  $\gamma$  is uniquely determined. In [19] the notation of "good" differential is introduced to distinguish those differentials which can happen with non-zero probability. A differential  $\alpha, \beta \rightarrow \gamma$  is not "good" iff for some  $i \in [0, n-1], \alpha_{i-1} = \beta_{i-1} = \gamma_{i-1} \neq \alpha_i \oplus \beta_i \oplus \gamma_i$  [19]. The exact probability of "good" differentials can be computed from Algorithm 2 of [19]. In general, it might not be easy to take redundancies into account. However, cryptanalysts should try their best. We also would like to emphasize that although there exists an exponential number of linear approximations (in terms of  $n_{\rm nl}$ ) for the compression function, it would be better in practice to concentrate on those for which highly probable linear differential paths are found easily. For example, by approximating the NUBFs with the zero function or sparse linear functions, the  $h \times m$  matrix  $\mathcal{H}$  which satisfies  $\mathsf{Compress}_{\mathsf{lin}}(\Delta) = \mathcal{H}\Delta$  is likely sparser, making it easier to find differential paths with good raw probability.

### 6 Application to MD6

MD6 [23], designed by Rivest et al., is a SHA-3 candidate that provides security proofs regarding some differential attacks. The core part of MD6 is the function f which works with 64-bit words and maps 89 input words  $(A_0, \ldots, A_{88})$  into 16 output words

 $(A_{16r+73},\ldots,A_{16r+88})$  for some integer r representing the number of rounds. Each round is composed of 16 steps. The function f is computed based on the following recursion

$$A_{i+89} = L_{r_i,l_i} (S_i \oplus A_i \oplus (A_{i+71} \wedge A_{i+68}) \oplus (A_{i+58} \wedge A_{i+22}) \oplus A_{i+72}),$$
 (9)

where  $S_i$ 's are some publicly known constants and  $L_{r_i,l_i}$ 's are some known simple linear transformations. The 89-word input of f is of the form Q||U||W||K||B where Q is a known 15-word constant value, U is a one-word node ID, W is a one-word control word, K is an 8-word key and B is a 64-word data block. For more details about function f and the mode of operation of MD6, we refer to the submission document  $[23]^6$ . We consider the compression function  $H = \operatorname{Compress}(M,V) = f(Q||U||W||K||B)$  where V = U||W||K, M = B and H is the 16-word compressed value. Our goal is to find a collision  $\operatorname{Compress}(M,V) = \operatorname{Compress}(M',V)$  for arbitrary value of V. We later explain how such collisions can be translated into collisions for the MD6 hash function.

According to our model (Section 5), MD6 can be implemented as an FSM which has  $64 \times 16r$  NUBFs of the form  $g(x,y,z,w) = x \cdot y \oplus z \cdot w$ . Remember that the NUBFs must not include any linear part or constant term. We focus on the case where we approximate all NUBFs with the zero function. This corresponds to ignoring the AND operations in equation (9). This essentially says that in order to compute Compress $_{\text{lin}}(\Delta) = \text{Compress}_{\text{lin}}(\Delta,0)$  for a 64-word  $\Delta = (\Delta_0,\ldots,\Delta_{63})$ , we map  $(A'_0,\ldots,A'_{24},A'_{25},\ldots,A'_{88}) = 0 || \Delta = (0,\ldots,0,\Delta_0,\ldots,\Delta_{63})$  into the 16 output words  $(A'_{16r+73},\ldots,A'_{16r+88})$  according to the linear recursion

$$A'_{i+89} = L_{r_i,l_i} (A'_i \oplus A'_{i+72}). \tag{10}$$

For a given  $\Delta$ , the function  $\Gamma$  is the concatenation of 16r words  $A'_{i+71} \vee A'_{i+68} \vee A'_{i+58} \vee A'_{i+22}$ ,  $0 \le i \le 16r-1$ . Therefore, the number of bit conditions equals

$$y = \sum_{i=0}^{16r-1} \operatorname{wt}(A'_{i+71} \vee A'_{i+68} \vee A'_{i+58} \vee A'_{i+22}).$$
 (11)

Note that this equation compactly integrates cases 1 and 2 given in section 6.9.3.2 of [23] for counting the number of active AND gates. Algorithm 3 in Appendix F shows how the Condition function is implemented using equations (6), (9) and (10).

Using a similar linear algebraic method to the one used in Section 4.4 for CubeHash, we have found the collision difference of equation (12) for r=16 rounds with a raw probability  $p_{\Delta}=2^{-90}$ . In other words,  $\Delta$  is in the kernel of Compress<sub>lin</sub> and the condition function has y=90 output bits. Note that this does not contradict the proven bound in [23]: one gets at least 26 active AND gates.

$$\Delta_{i} = \begin{cases} \text{F6D164597089C40E} & i = 2\\ 2000000000000000 & i = 36\\ 0 & 0 \le i \le 63, i \ne 2, 36 \end{cases}$$
 (12)

**Table 6.** Input and output partitionings of the Condition function of MD6 with r=16 rounds.

i	${\cal M}_i$	$\mathcal{Y}_i$	$q_i$	$ q_i' $
0	-	Ø	0	0
1	$\{M_{38}\}$	$\{Y_1, \ldots, Y_{29}\}$	29	0
2	$\{M_{55}\}$	$\{Y_{43},\ldots,Y_{48}\}$	6	0
3	$\{M_0, M_5, M_{46}, M_{52}, M_{54}\}$	$\{Y_0\}$	1	0
4	$\{M_j   j = 3, 4, 6, 9, 21, 36, 39, 40, 42, 45, 49, 50, 53, 56, 57\}$	$\{Y_{31},\ldots,Y_{36}\}$	6	0
5	$\{M_{41}, M_{51}, M_{58}, M_{59}, M_{60}\}$	$\{Y_{30}, Y_{51}\}$	2	0
6	$ \{M_j j=1,2,7,8,10,11,12,17,18,20,22,24,25,26,29,$	$\{Y_{52},\ldots,Y_{57}\}$	6	0
	33, 34, 37, 43, 44, 47, 48, 61, 62, 63			1
7	$\{M_{27}\}$	$\{Y_{37},\ldots,Y_{42}\}$	6	0
8	$\{M_{13}, M_{16}, M_{23}\}$	$\{Y_{50}\}$	1	0
9	$\{M_{35}\}$	$\{Y_{49}\}$	1	0
10	$\{M_{14}, M_{15}, M_{19}, M_{28}\}$	$\{Y_{58}, Y_{61}\}$	2	0
11	$\{M_{30}, M_{31}, M_{32}\}$	$\{Y_{59}, Y_{60}, Y_{62} \dots, Y_{89}\}$	30	0

In order to efficiently find a conforming message pair for this differential path we need to analyze the dependency table of its condition function. Referring to our notations in Section 3.2, our analysis of the dependency table of function Condition $_{\Delta}(M,0)$  at word level (units of u=64 bits) shows that the partitioning of the condition function is as in Table 6 for threshold value th=0. For this threshold value clearly  $p_i=0$ . The optimal values for  $q_i$ 's (computed according to the complexity analysis of the same section) are also given in Table 6, showing a total attack complexity of  $2^{30.6}$  (partial) condition function evaluation<sup>7</sup>. By analyzing the dependency table with smaller units the complexity may be subject to reduction.

Having set V to zero (which corresponds to choosing null values for the key, the node ID and the control word in order to simplify things), we found a message M, given in the Appendix E, which makes the condition function null. In other words, the message pairs M and  $M \oplus \Delta$  are colliding pairs for r=16 rounds of f. This 16-round colliding pair provides near collisions for r=17, 18 and 19 rounds, respectively, with 63, 144 and 270 bit differences over the 1024-bit long output of f.

Now, let's discuss the potential of providing collisions for reduced-round MD6 hash function from collisions for reduced-round f. The MD6 mode of operation is optionally parametrized by an integer L,  $0 \le L \le 64$ , which allows a smooth transition from the default tree-based hierarchical mode of operation (for L=64) down to an iterative mode of operation (for L=0). When L=0, MD6 works in a manner similar to that of the well-known Merkle-Damgård construction (or the HAIFA method). Since in the iterative Merkle-Damgård the first 16 words of the message block are used as a chaining value, and as our difference in equation (12) is non-zero in the first 16 words, we do not get a collision but a pseudo-collision. Nevertheless, for 16-round MD6 in the tree-based hierarchical mode of operation (i.e. for  $1 \le L \le 64$ ) we get a hash collision. We emphasize that one must choose node ID U and control word W properly in order

<sup>&</sup>lt;sup>6</sup> In the MD6 document [23], the control word C and the linear function  $L_{r_i,l_i}$  are respectively denoted by V and  $g_{r_i,l_i}$ .

 $<sup>^7</sup>$  By masking  $M_{38}$  and  $M_{55}$  respectively with 092E9BA68F763BF1 and DFFBFF7FEFFDFFBF after random setting, the 35 condition bits of the first three steps are satisfied for free, reducing the complexity to  $2^{30.0}$  instead.

to fulfill the MD6 restriction on these values as opposed to the null values which we chose. This is the first real collision example for 16-round MD6. The original MD6 submission [23] mentions inversion of the function f up to a dozen rounds using SAT solvers. Some slight nonrandom behavior of the function f up to 33 rounds has also been reported [17].

### 7 Conclusion

We presented a framework for an in-depth study of linear differential attacks on hash functions. We applied our method to reduced round variants of CubeHash and MD6, giving by far the best known collision attacks on these SHA-3 candidates. Our results may be improved by considering start-in-the middle attacks if the attacker is allowed to choose the initial value of the internal state.

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# A Highest raw probability differential paths found for CubeHash

Here we give differential trials for highlighted entries of Table 2. Remember that  $\Delta = \Delta^0 ||\Delta^1||...||\Delta^{t-1}$  is in the kernel of  $\mathsf{Compress}_{\mathsf{lin}}$ , the linearized version of  $\mathsf{Compress}_{\mathsf{lin}}$  Compress $_{\mathsf{lin}}$  linearly maps these t block differences (each of b bytes) into the last 128-b bytes of the final state. The erasing block difference  $\Delta^t$  is then the difference in the first b bytes of the state, hence only depending on  $\Delta$ . See sections 4.2 and 4.4 for more details. Note that a differential path for a given r and b can be easily transformed to a differential path for the same r and bigger b's by appending enough zeros to each difference block. Therefore we introduce the differential path for the smallest valid b.

### A.1 Differential paths for CubeHash-1/\*

CubeHash-1/4 with y = 32:

$$(\Delta^0, \dots, \Delta^4) = (00000001, 00000000, 40400010, 00000000, 00000010)$$

CubeHash-1/3 with y = 46:

$$(\Delta^0, \dots, \Delta^4) = (010000, 000000, 104040, 000000, 100000)$$

CubeHash-1/2 with y = 221:

$$(\Delta^0, \dots, \Delta^8) = (0080, 0000, 2200, 0000, 228A, 0000, 0280, 0000, 2000)$$

### A.2 Differential paths for CubeHash-2/\*

CubeHash-2/4 with y = 32:

$$(\Delta^0, \Delta^1, \Delta^2) = (00000001, 40400010, 00000010)$$

CubeHash-2/3 with y = 46:

$$(\Delta^0, \Delta^1, \Delta^2) = (010000, 104040, 100000)$$

CubeHash-2/2 with y = 221:

$$(\Delta^0, \dots, \Delta^4) = (0040, 1100, 1145, 0140, 1000)$$

# A.3 Differential paths for CubeHash-3/\*

CubeHash-3/64 with y = 65:

#### CubeHash-3/48 with y = 364:

# CubeHash-3/12 with y = 400:

 $\begin{array}{l} \varDelta^0 = \texttt{000200000000000000020000} \\ \varDelta^1 = \texttt{0000000000000000000000000} \\ \varDelta^2 = \texttt{800A2A820000000000000000} \\ \varDelta^3 = \texttt{00000000000000000000000000000} \\ \varDelta^4 = \texttt{000020000000000000000000000000000} \end{array}$ 

CubeHash-3/4 with y = 478:

 $(\Delta^0, \dots, \Delta^4) = (00000100, 00000000, 41400515, 00000000, 00000010)$ 

CubeHash-3/3 with y = 798:

 $(\Delta^0, \dots, \Delta^4) = (000280, 000000, 0A2A8A, 000000, 080020)$ 

# A.4 Differential paths for CubeHash-4/\*

CubeHash-4/36 with y = 130:

CubeHash-4/12 with y = 156:

 $\Delta^0 = \texttt{04000000000000004000000}$ 

 $\varDelta^1 = \texttt{00440040000000000440040}$ 

 $\Delta^2 = 000400000000000000040000$ 

CubeHash-4/4 with y = 189:

$$(\Delta^0, \Delta^1, \Delta^2) = (00001000, 01000110, 00000010)$$

CubeHash-4/3 with y = 195:

$$(\Delta^0, \Delta^1, \Delta^2) = (000200, 200022, 000002)$$

# A.5 Differential paths for CubeHash-5/\*

CubeHash-5/64 with y = 205:

### A.6 Differential paths for CubeHash-6/\*

CubeHash-6/36 with y = 351:

CubeHash-6/12 with y = 400:

CubeHash-6/4 with y = 478:

$$(\Delta^0, \Delta^1, \Delta^2) = (00000100, 41400515, 00000010)$$

### A.7 Differential paths for CubeHash-7/\*

CubeHash-7/64 with y = 447:

# B Best total attack complexity differential paths found for CubeHash

Here we give the differential trials for some stared entries of Table 3. Please see the note on Appendix A.

CubeHash-3/48 with y = 368:

## CubeHash-4/36 with y = 134:

### CubeHash-7/61 with y = 455:

# C Collisions for CubeHash-3/64

- $M^{\rm pre} = 9$ B91E97363511AC3AF950F54DBCFD5DF91BC26BDD759104D F15B37847A4F7015E15A8844ABA3075A3816AE13E583F276 40193317724464649F9BE819EB582ECC
- $M^0 = \mbox{ B22A98139CC0C8606525818EE6DD7775CF25B34196DC51F4} \\ 641E56ACB918296BBD082AD01D7481EECC950B6C176C45B6 \\ 23CFE1E2638B16255F61E806F34DE91C$
- $\begin{array}{ll} M^1 = & \texttt{4D9E9CD62ED12CBDBA1E0B631856DCFE5BD996571CFF6E94} \\ & \texttt{A52242382E154FA6AEB44AC0A247CB298550C7B82BDCA924} \\ & \texttt{E81D5E51E997CA67FBDD86FF15D04A0D} \end{array}$

### D Collisions for CubeHash-4/48

- $M^{\rm pre} = 741 \text{B} 87597 \text{F} 94 \text{FF} 1\text{CC} 01761 \text{CA} 0\text{D} 80 \text{B} 07 \text{CC} 2\text{E} 6\text{E} 760 \text{C} 95 \text{D} \text{F} 9\text{A} 5$  08 FF CB AB DA 11474 E 2CCE A 7AC 62 A 7C 822 BE 29 ED CB A 99 D 476 C
- $M^0 = 1$ D30F8022F4AE8DBD477FA1F7DE37C1AF2516BC6FA4657F9 E51539C10EC114DA3B8264DD9361FE07C3D56E88E8512201
- $M^1 = 014$ A11BFE2FF346FC306D1E430EE80268785A9F841562C9A 88A6BF5858E95362F541ACF41C2FDCC1C49470DF1DFAEFDC

# E Colliding message for MD6 reduced to 16 rounds

```
5361E9B8579F7CD1, 8B29C52CA2AB51E4, 0BCF2F1E1B116898, 022C254B88191A11,
F0F1CE9D9A7F63B8, 9FB5B2CE87B7D7F5, E7C78F28EEB4F5C7, C5E8C19CEFC07365,
F88B84529ED90209, 8FACF593AE7390CF, 03A93466247C6B54, B12C70C10904143D,
D92EE67244C300D6, 35EEA586ECCC8A77, 9DCF031C64B528F8, C84807607ADCD418,
367E95EE3CB0FC67, 578A2C716FCC5016, B0C30EA5521F61EF, 7F665B24762D5894,
4196BAF0596A7784, ED5F9A8F183B4BCC, 6077463601FCFE46, 495366B1273E119B,
6E11A21AE5B3A48F, 38082264A0F68F93, 4ED510C2DFA9FF98, 35C5ACEC5E9A1756,
1F6731C861879ECD, 8CECD7B4F761CE82, 332A50854FDA8FE6, 588498B1021E9C23,
CB1FFA21CF89C7A5, 63A6871C77848410, 92A550CB4607F31C, 97024803F162E055,
E2D6EA5A57D2DBF3, AEB418A0F1F01CC5, 090A9304040038C1, 5417960E3D9A06A5,
714215C196813F35, BABAD7A4C154F2C3, 71AF3FD02B543940, FA08624B825648DD,
730D61FF48759275, CF85BA5A06D6AED4, 2E12B3150452C65A, 93C7A9FC314220B4,
81B128A4EF361456, BFE652098170C212, 77540989DC246845, 796F353D07721071,
D82776A3CBFEC586, 1132E4391152F408, CE936924CFFB22AA, D338852F80450282,
4F41AB82E790EEF6, F05378CB6BD36203, 5E506F47C6EC4617, FE6FB5A03BDE8E1C,
AB33EA511EEBAEDC, 7D40F8D4F0C62BF4, 1174E2B748B9CC2E, 1EB743671A31547D
```

# F Condition function for CubeHash and MD6

Algorithms 3 and 4 respectively show how the Condition function can be constructed for MD6 and CubeHash. It is presumed that the input  $\Delta$  is in the kernel of their respective condition functions. Note that, in case the Condition function needs to be evaluated several times for a fixed differential path  $\Delta$ , as is the case for the tree-based search algorithm, we can precompute that part of the function which is independent of the input message M.

### **Algorithm 3**: Condition function for MD6

```
Require: r, \Delta, M and V
Ensure: y, Y = \mathsf{Condition}_{\Delta}(M, V)
 1: (A_0, \ldots, A_{88}) := Q||V||M
 2: (A'_0, \dots, A'_{24}, A'_{25}, \dots, A'_{88}) := 0 || \Delta
 3: i := 0
 4: for i = 0, 1, \dots, 16r - 1 do
         A_{i+89} := L_i(S_i \oplus A_i \oplus (A_{i+71} \wedge A_{i+68}) \oplus (A_{i+58} \wedge A_{i+22}) \oplus A_{i+72})
         A'_{i+89} := L_i (A'_i \oplus A'_{i+72})
D := A'_{i+71} \vee A'_{i+68} \vee A'_{i+58} \vee A'_{i+22}
 7:
         T := (A_{i+71} \wedge A_{i+68}) \oplus (A_{i+58} \wedge A_{i+22})
 9:
         T' := ((A_{i+71} \oplus A'_{i+71}) \land (A_{i+68} \oplus A'_{i+68})) \oplus ((A_{i+58} \oplus A'_{i+58}) \land (A_{i+22} \oplus A'_{i+58}))
         for all bit positions k = 0, 1, \dots, 63 such that D_k = 1 do
10:
11:
             Y_j := T_k \oplus T'_k
12:
             j := j + 1
         end for
13:
14: end for
15: y := j
```

### Algorithm 4: Condition function for CubeHash

```
Require: r, b, t, \Delta = \Delta^0 || \dots || \Delta^{t-1}, M = M^0 || \dots || M^{t-1} \text{ and } V = (V_0, \dots, V_{31})
Ensure: y, Y = Condition_{\Delta}(M, V)
 1: (S_0, \ldots, S_{31}) := V
 2: (S'_0, \dots, S'_{31}) := (0, \dots, 0)
 3: j := 0
 4: for t' from 0 to t-1 do
         XOR M^{t'} into the first b bytes of the state S {see footnote 5}
         XOR \Delta^{t'} into the first b bytes of the state S' {see footnote 5}
 6:
         for round from 1 to r do
 7:
             for i from 0 to 15 do
 8:
 9:
                \alpha := S'_i, \beta := S'_{i \oplus 16}, A := S_i, B := S_{i \oplus 16}
10:
                Add S_i into S_{i\oplus 16} but XOR S'_i into S'_{i\oplus 16}
11:
                D := \alpha \vee \beta
                C := A \oplus B \oplus S_{i \oplus 16}  {carry word}
12:
                T = ((\alpha \oplus \beta) \land C) \oplus (\alpha \land B) \oplus (\beta \land A) \oplus (\alpha \land \beta)  {see equation (5) or (8)}
13:
                for all bit positions k = 0, 1, ..., 31 such that D_k = 1 do
14:
15:
                    Y_j := T_k
                    j := j + 1
16:
                end for
17:
18:
             end for
             for 0 \le i \le 15 do rotate S_i and S'_i to the left by seven bits end for
19:
20:
             for 0 \le i \le 7 do swap S_i and S_{i \oplus 8} as well as S'_i and S'_{i \oplus 8} end for
             for 0 \le i \le 15 do XOR S_{i \oplus 16} into S_i and S'_{i \oplus 16} into S'_i end for
21:
22:
             for i \in \{16, 17, 20, 21, 24, 25, 28, 29\} do swap S_i and S_{i\oplus 2} as well as S_i' and S_{i\oplus 2}'
             end for
              \  \, \textbf{for} \ i \ \mathsf{from} \ 0 \ \mathsf{to} \ 15 \ \ \textbf{do} \\
23:
24:
                \alpha := S'_i, \beta := S'_{i \oplus 16}, A := S_i, B := S_{i \oplus 16}
                Add S_i into S_{i\oplus 16} but XOR S_i' into S_{i\oplus 16}'
25:
26:
                D := \alpha \vee \beta
                C := A \oplus B \oplus S_{i \oplus 16} \{ \text{carry word} \}
27:
28:
                T = ((\alpha \oplus \beta) \land C) \oplus (\alpha \land B) \oplus (\beta \land A) \oplus (\alpha \land \beta)  {see equation (5) or (8)}
                for all bit positions k = 0, 1, ..., 31 such that D_k = 1 do
29:
                    Y_i := T_k
30:
                    j := j + 1
31:
32:
                end for
33:
34:
             for 0 \le i \le 15 do rotate S_i and S_i' to the left by eleven bits end for
35:
             for i \in \{0, 1, 2, 3, 8, 9, 10, 11\} do swap S_i and S_{i \oplus 4} as well as S_i' and S_{i \oplus 4}' end for
36:
             for 0 \le i \le 15 \text{ XOR } S_{i \oplus 16} do into S_i and S'_{i \oplus 16} into S'_i end for
             for i \in \{16, 18, 20, 22, 24, 26, 28, 30\} do swap S_i and S_{i\oplus 1} as well as S_i' and S_{i\oplus 1}'
37:
             end for
         end for \{r \text{ rounds}\}
39: end for\{t \text{ iterations}\}
40: y := j
```