Efficient Dissection of Composite Problems, with Applications to Cryptanalysis, Knapsacks, and Combinatorial Search Problems

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Abstract. In this paper we show that a large class of diverse problems have a bicomposite structure which makes it possible to solve them with a new type of algorithm called dissection, which has much better time/memory tradeoffs than previously known algorithms. A typical example is the problem of finding the key of multiple encryption schemes with r independent n-bit keys. All the previous error-free attacks required time T and memory M satisfying $TM = 2^{rn}$, and even if "false negatives" are allowed, no attack could achieve $TM < 2^{3rn/4}$. Our new technique yields the first algorithm which never errs and finds all the possible keys with a smaller product of TM, such as $T=2^{4n}$ time and $M=2^n$ memory for breaking the sequential execution of r=7 block ciphers. The improvement ratio we obtain increases in an unbounded way as r increases, and if we allow algorithms which can sometimes miss solutions, we can get even better tradeoffs by combining our dissection technique with parallel collision search. To demonstrate the generality of the new dissection technique, we show how to use it in a generic way in order to attack hash functions with a rebound attack, to solve hard knapsack problems, and to find the shortest solution to a generalized version of Rubik's cube with better time complexities (for small memory complexities) than the best previously known algorithms.

Keywords: Cryptanalysis, TM-tradeoff, multi-encryption, knapsacks, bicomposite, dissection, rebound

1 Introduction

A composite problem is a problem that can be split into several simpler subproblems which can be solved independently of each other. To prevent attacks based on such decompositions, designers of cryptographic schemes usually try to entangle the various parts of the scheme by using a complex key schedule in block ciphers, or a strong message expansion in hash functions. While we can formally split such a structure into a top part that processes the input and a bottom part that produces the output, we cannot solve these subproblems independently of each other due to their strong interactions.

However, when we deal with higher level constructions which combine multiple primitives as black boxes, we often encounter unrelated keys or independently computed outputs which can provide exploitable decompositions. One of the best examples of such a situation was the surprising discovery by Joux [7] in 2004 that finding collisions in hash functions defined by the *parallel execution* of several independent hash functions is much easier than previously believed. In this paper we show the dual result that finding the key of a multiple-encryption scheme defined by the *sequential execution* of several independent cryptosystems is also easier than previously believed.

Since we can usually reduce the time complexity of cryptanalytic attacks by increasing their memory complexity, we will be interested in the full tradeoff curve between these two complexities rather than in a single point on it. We will be primarily interested in algorithms which use an exponential combination of $M=2^{mn}$ memory and $T=2^{tn}$ time for a small constant m and a larger constant t, when the key size n grows to infinity. While this setup may sound superficially similar to Hellman's time/memory tradeoff algorithms, it is important to notice that Hellman's preprocessing phase requires time which is equivalent to exhaustive search and memory which is at least the square root of the number of keys, and that in Hellman's online phase the product of time and memory is larger than the number of keys. In our model we do not allow free preprocessing, we can use smaller amounts of memory, and the product of time and memory is strictly smaller than the number of keys.

The type of problems we can solve with our new techniques is characterized by the existence of two orthogonal ways in which we can decompose a given problem into (almost) independent parts. We call such problems bicomposite, and demonstrate this notion by considering the problem of cryptanalyzing the sequential execution of r block ciphers which use independent n-bit keys to process n-bit plaintexts. In order to make the full rn-bit key of this scheme unique with a reasonable probability, the cryptanalyst needs r known plaintext/ciphertext pairs. The full encryption process can thus be described by an $r \times r$ matrix whose columns corresponds to the processing of the various plaintexts and whose rows correspond to the application of the various block ciphers. The attacker is given the r plaintexts at the top and the r ciphertexts at the bottom, and his goal is to find all the keys with a generic algorithm which does not assume the existence of any weaknesses in the underlying block ciphers. The reason we say that this problem is bicomposite is that the keys are independently chosen and the plaintexts are independently processed, and thus we can partition the execution matrix both horizontally and vertically into independent parts. In particular, if we know certain subsets of keys and certain subsets of intermediate values, we can independently verify their consistency with the given plaintexts or ciphertexts without knowing all the other values in the execution matrix. This should be contrasted with the standard constructions of iterated block ciphers, in which a partial guess of the key and a partial guess of some state bits in the middle of the encryption process cannot be independently verified by an efficient computation.

The security of multiple-encryption schemes had been analyzed for more than 30 years, but most of the published papers had dealt with either double or triple encryption (which are widely used as DES-extensions in the banking industry). While the exact security of double and triple encryption are well understood and we can not push their analysis any further, our new techniques show that surprisingly efficient attacks can be applied already when we make the next step and consider quadruple encryption, and that additional improvements can be made when we consider even longer combinations.

Standard meet-in-the-middle (MITM) attacks, which account for the best known results against double and triple encryption, try to split such an execution matrix into a top part and a bottom part with a single horizontal partition line which crosses the whole matrix from left to right. Our new techniques use a more complicated way to split the matrix into independent parts by exploiting its two dimensional structure. Consider, for example, the sequential execution of 7 independent block ciphers. We can find the full 7n-bit key in just 2^{4n} time and 2^n memory by guessing two of the seven internal states after the application of the third block cipher and one of the seven internal states after the application of the fifth block cipher. We call such an irregular way to partition the execution matrix with partial guesses a dissection, since it mimics the way a surgeon operates on a patient by using multiple cuts of various lengths at various locations.

Our new techniques make almost no assumptions about the internal structure of the primitive operations, and in particular they can be extended with just a slight loss of efficiency to primitive operations which are one-way functions rather than easily invertible permutations. This makes it possible to find improved attacks on message authentication codes (MACs) which are defined by the sequential execution of several keyed hash functions. Note that standard MITM attacks cannot be applied in this case, since we have to encrypt the inputs and decrypt the outputs in order to compare the results in the middle of the computation.

To demonstrate the generality of our techniques, we show in this paper how to apply them to several types of combinatorial search problems (some of which have nothing to do with cryptography), such as the knapsack problem: Given n generators a_1, a_2, \ldots, a_n which are n-bit numbers, find a subset that sums modulo 2^n to S. The best known special purpose algorithm for this problem was published at Eurocrypt 2011 by Becker et al. [1], but our generic dissection technique provides better time complexities for small memory complexities. To show the connection between knapsack problems and multiple-encryption, describe the solution of the given knapsack problem as a two dimensional $r \times r$ execution matrix, in which we partition the generators into r groups of n/r generators, and partition each number into r blocks of r0 generators from the matrix is defined by adding the appropriate subset of generators from the next group to the accumulated sum computed in the previous row. We start with an initial value of zero, and our problem is to find some execution that leads

to a desired value S after the last row. This representation is bicomposite since the choices made in the various rows of this matrix are completely independent, and the computations made in the various columns of this matrix are almost independent since the only way they interact with each other is via the addition carries which do not tend to propagate very far into the next block. This makes it possible to guess and operate on partial states, and thus we can apply almost the same dissection technique we used for multiple-encryption schemes. Note that unlike the case of multiple-encryption in which the value of r was specified as part of the given problem, here we can choose any desired value of r independently of the given value of r in order to optimize the time complexity for any available amount of memory. In particular, by choosing r=7 we can reduce the best known time complexity for hard knapsacks when we use $M=2^{n/7}=2^{0.1428n}$ memory from $2^{(3/4-1/7)n}=2^{0.6071n}$ in [1] to $2^{4n/7}=2^{0.5714n}$ with our new algorithm.

The algorithm of Becker et al. [1] crucially depends on the fact that addition is an associative and commutative operation on numbers, and that sets can be partitioned into the union of two subsets in an exponential number of ways. Our algorithms make no such assumptions, and thus they can be applied under a much broader set of circumstances. For example, consider a non-commutative variant of the knapsack problem in which the generators a_i are permutations over $\{1, 2, \dots, k\}$, and we have to find a product of length ℓ of these generators which is equal to some given permutation S (a special case of this variant is the problem of finding the fastest way to solve a given state of Rubik's cube by a sequence of face rotations, which was analyzed extensively in the literature). To show that this problem is bicomposite, we have to represent it by an execution matrix with independent rows and columns. Consider an $\ell \times k$ matrix in which the i-th row represents the action of the i-th permutation in the product, and the j-th column represents the current location of element j from the set. Our goal is to start from the identity permutation at the top, and end with the desired permutation S at the bottom. We can reduce this matrix to size $r \times r$ for any desired r by bunching together several permutations in the product and several elements from the set. The independence of the rows in this matrix follows from the fact that we can freely choose the next generators to apply to the current state, and the independence of the columns follows from the fact that we can know the new location of each element j if we know its previous location and which permutation was applied to the state, even when we know nothing about the locations of the other elements in the previous state. This makes it possible to guess partial states at intermediate stages, and thus to apply the same dissection algorithms as in the knapsack problem with the same improved complexities.

We note that generic ideas similar to the basic dissection attacks were used before, in the context of several specific bicomposite problems. These include the algorithms of Schroeppel and Shamir [14] and of Becker et al. [1] which analyzed the knapsack problem, the algorithm of van Oorschot and Wiener [15] which attacked double and triple encryption, and the results of Dinur et al. [3] in the specific case of the block cipher GOST. A common feature of all these

algorithms is that none of them could beat the tradeoff curve $TM = N^{3/4}$, where N is the total number of keys. The algorithms of [3,14,15] matched this curve only for a single point, and the recent algorithm of Becker et al. [1] managed to match it for a significant portion of the tradeoff curve. Our new dissection algorithms not only allow to beat this curve, but actually allow to obtain the relation $TM < N^{3/4}$ for any amount of memory in the range $M \le N^{1/4}$.

The paper is organized as follows: In Section 3 we introduce the dissection technique and present our best error-free attacks on multiple encryption. In Section 4 we consider the model when "false negatives" are allowed, and show that the dissection algorithms can be combined with the parallel collision algorithm of van Oorschot and Wiener [15] to get an improved time-memory tradeoff curve. Finally, in Section 5 we apply our results to various problems, including knapsacks, rebound attacks on hash functions and search problems in databases.

2 Notations and Conventions

In this paper we denote the basic block cipher by E and assume that it uses n-bit blocks and n-bit keys (we can easily deal with other sizes, but it makes the notation cumbersome). We denote by E^i the encryption process with key k_i , and denote by $E^{[1...r]}$ the multiple-encryption scheme which uses r independent keys to encrypt the plaintext P and produce the ciphertext C via $C = E_{k_r}(E_{k_{r-1}}(\cdots E_{k_2}(E_{k_1}(P))\cdots))$. The intermediate value produced by the encryption of P under $E^{[1...r]}$ is denoted by X^i , and the decryption process of $E^{[1...r]}$ is denoted by $D^{[1...r]}$ (which applies the keys in reverse order). To attack $E^{[1...r]}$, we are usually given r plaintext/ciphertext pairs, which are expected to make the key unique (at intermediate stages, we may be given fewer than i-j+1 plaintext/ciphertext pairs for $E^{[i...j]}$, and then we are expected to produce all the compatible keys). In all our exponential complexity estimates, we consider expected rather than maximal possible values (under standard randomness assumptions, they differ by no more than a logarithmic factors), and ignore multiplicative polynomial factors in n and r.

3 Dissecting the Multiple-Encryption Problem

In this section we develop our basic dissection algorithms that allow to solve efficiently the problem of multiple encryption. Given r-encryption with r independent keys, r n-bit plaintext/ciphertext pairs and 2^{mn} memory cells, the algorithms find all possible values of the keys which comply with the plaintext/ciphertext pairs, or prove that there are no such keys. The algorithms are deterministic, in the sense that they do not need random bits and they always succeed since they implicitly scan all possible solutions.

First, we treat the case of general r and m = 1. We then generalize the algorithms to other integer values of m (in Appendix A.2). The algorithms can be extended also to fractional values of m and to compositions of one-way functions,

which appear in the context of layered Message Authentication Codes, such as NMAC [2]. The first non-integer extension is presented in the full version of the paper, whereas the extension to one-way functions is presented in Appendix A.4.

3.1 Previous Work — The Meet in the Middle Attack

The trivial algorithm for recovering the key of r-encryption is exhaustive search over the 2^{rn} possible key values, whose time complexity is 2^{rn} , and whose memory requirement is negligible. In general, with no additional assumptions on the algorithm and on the subkeys, this is the best possible algorithm.

In [12] Merkle and Hellman observed that if the keys used in the encryption are independent, an adversary can trade time and memory complexities, using a meet in the middle approach. In this attack, the adversary chooses a value u, $1 \le u \le \lfloor r/2 \rfloor$, and for each possible combination of the first u keys $(k_1, k_2, \dots k_u)$ she computes the vector $(X_1^u, X_2^u, \dots, X_r^u) = E^{[1 \dots u]}(P_1, P_2, \dots, P_r)$ and stores it in a table (along with the respective key candidate). Then, for each value of the last r-u keys, the adversary computes the vector $D^{[u+1 \dots r]}(C_1, C_2, \dots, C_r)$ and checks whether the value appears in the table (each such collision suggests a key candidate (k_1, \dots, k_r)). The right key is necessarily suggested by this approach, and in cases when other keys are suggested, additional plaintext/ciphertext pairs can be used to sieve the wrong key candidates.

The time complexity of this algorithm is $T = 2^{(r-u)n}$, whereas its memory complexity is $M = 2^{un}$. Hence, the algorithm allows to achieve the tradeoff curve $TM = 2^{rn}$ for any values T, M such that $M \leq 2^{\lfloor r/2 \rfloor n}$. Note that the algorithm can be applied also if the number of available plaintext/ciphertext pairs is r' < r. In such case, it outputs all the possible key candidates, whose expected number is $2^{(r-r')n}$ (since the plaintext/ciphertext pairs yield an r'n-bit condition on the 2^{rn} possible keys).

The meet in the middle attack, designed for breaking double-encryption, is still the best known generic attack on double encryption schemes. It is also the best known attack for triple encryption upto logarithmic factors,² which was studied very extensively due to its relevance to the former de-facto encryption standard Triple-DES.

3.2 The Basic Dissection Algorithm: Attacking 4-Encryption

In the followings we show that for $r \geq 4$, the basic meet in the middle algorithm can be outperformed significantly, using a dissection technique. For the basic

¹ We note that the algorithm, as described above, works only for $u \in \mathbb{N}$. However, it can be easily adapted to non-integer values of $u \leq \lfloor r/2 \rfloor$, preserving the tradeoff curve $TM = 2^{rn}$.

² We note that a logarithmic time complexity improvement can be achieved in these settings as suggested by Lucks [10]. The improvement relies on the variance in the number of keys encrypting a given plaintext to a given ciphertext. This logarithmic gain in time complexity comes hand in hand with an exponential increase in the data complexity (a factor 8 gain in the time complexity when attacking triple-DES increases the data from 3 plaintext-ciphertext pairs to 2⁴⁵ such pairs).

case r = 4, considered in this section, our algorithm runs in time $T = 2^{2n}$ with memory 2^n , thus allowing to reach $TM = 2^{3n}$, which is significantly better than the $TM = 2^{4n}$ curve suggested by the meet-in-the-middle attack.

The main idea behind the algorithm is to dissect the 4-encryption into two 2-encryption schemes, and to apply the meet in the middle attack to each of them separately. The partition is achieved by enumerating parts of the internal state at the dissection point. The basic algorithm, which we call $Dissect_2(4,1)$ for reasons which will become apparent later, is as follows:

- 1. Given plaintexts (P_1, P_2, P_3, P_4) and their corresponding ciphertexts (C_1, C_2, C_3, C_4) , for each candidate value of $X_1^2 = E_{k_2}(E_{k_1}(P_1))$:
- 2. (a) Run the standard meet in the middle attack on 2-round encryption with (P_1, X_1^2) as a single plaintext-ciphertext pair. For each of the 2^n values of (k_1, k_2) output by the attack, partially encrypt P_2 using (k_1, k_2) , and store in a table the corresponding values of X_2^2 , along with the values of (k_1, k_2) .
 - (b) Run the standard meet in the middle attack on 2-round encryption with (X_1^2, C_1) as a single plaintext-ciphertext pair. For each of the 2^n values of (k_3, k_4) , partially decrypt C_2 using (k_3, k_4) and check whether the suggested value for X_2^2 appears in the table. If so, check whether the key (k_1, k_2, k_3, k_4) suggested by the table and the current (k_3, k_4) candidate encrypts P_3 and P_4 into C_3 and C_4 , respectively.

It is easy to see that once the right value for X_1^2 is considered, the right values of (k_1, k_2) are found in Step 2(a) and the right values of (k_3, k_4) are found in Step 2(b), and thus, the right value of the key is necessarily found. The time complexity of the algorithm is 2^{2n} . Indeed, Steps 2(a) and 2(b) are called 2^n times (for each value of X_1^2), and each of them runs the basic meet in the middle attack on 2-encryption in expected time and memory of 2^n . The number of expected collisions in the table of X_2^2 is 2^n . Thus, the expected time complexity of the attack³ is $2^n \cdot 2^n = 2^{2n}$.

The memory consumption of the 2-encryption meet in the middle steps is expected to be about 2^n . The size of the table "passed" between Steps 2(a) and 2(b) is also 2^n , since each meet in the middle step is expected to output 2^n key candidates. Hence, the expected memory complexity of the entire algorithm is 2^n .

3.3 Natural Extensions of the Basic Dissection Algorithm

We now consider the case (r > 4, m = 1) and show that natural extensions of the $Dissect_2(4,1)$ algorithm presented above, allow to increase the gain over the standard meet in the middle attack significantly for larger values of r.

It is clear that any algorithm for r'-encryption can be extended to attack r-encryption for any r > r', by trying all possible r - r' keys $(k_{r'+1}, \ldots, k_r)$, and applying the basic algorithm to the remaining $E^{[1...r']}$. The time complexity is

³ We remind the reader that we disregard factors which are polynomial in n and r.

increased by a multiplicative factor of $2^{(r-r')n}$, and hence, the ratio $2^{rn}/TM$ is preserved. This leads to the following natural definition.

Definition 1. The gain of an algorithm A for r-encryption whose time and memory complexities are T and M, respectively, is defined as $Gain(A) = \log(2^{rn}/TM)/n = r - \log(TM)/n$. The maximal gain amongst all deterministic algorithms for r-encryption which use 2^{mn} memory, is denoted by $Gain_D(r,m)$.

By the trivial argument above, $Gain_D(r, 1)$ is monotone non-decreasing with r. The $Dissect_2(4, 1)$ algorithm shows that $Gain_D(r, 1) \ge 1$ for r = 4, and hence, for all $r \ge 4$. Below we suggest two natural extensions, which allow to increase the gain up to \sqrt{r} .

The LogLayer Algorithm: The first extension of the $Dissect_2(4,1)$ is the recursive $LogLayer_r$ algorithm, applicable when r is a power of 2, which tries all the possible X_1^{2i} for $i=1,2,\ldots,r/2-1$ and runs simple meet in the middle attacks on each subcipher $E^{[2i+1\ldots 2i+2]}$ separately. As each such attack returns 2^n candidate keys (which can be stored in memory of $(r/2) \cdot 2^n$), the algorithm then groups 4 encryptions together, enumerates the values X_2^{4i} for $i=1,2,\ldots,r/4-1$, and runs meet in the middle attacks on each subcipher $E^{[4i+1\ldots 4i+4]}$ separately (taking into account that there are only 2^n possibilities for the keys (k_{4i+1}, k_{4i+2}) and 2^n possibilities for the keys (k_{4i+3}, k_{4i+4})). The algorithm continues recursively (with $\log r$ layers in total), until a single key candidate is found.

The memory complexity of $LogLayer_r$ is 2^n (as we need to store a number linear in r of lists of 2^n keys each). As in the j-th layer of the attack, $(r/2^j) - 1$ intermediate values are enumerated, and as each basic meet in the middle attack has time complexity of 2^n , the overall time complexity of the attack is

$$\prod_{j=1}^{\log r} 2^{n((r/2^j)-1)} \cdot 2^n = 2^{n(r-\log r)}.$$

Therefore, the gain is $Gain(LogLayer_r) = \log r - 1$, which shows that $Gain_D(r, 1) \ge \lfloor \log r \rfloor - 1$.

The $Square_r$ Algorithm: This logarithmic gain of LogLayer can be significantly outperformed by the $Square_r$ algorithm, applicable when $r=(r')^2$ is a perfect square. The $Square_r$ algorithm starts by trying all the possible values of r'-1 intermediate values every r' rounds, a total of $(r'-1)^2$ intermediate encryption values. Namely, the algorithm starts by enumerating all $X_1^{r'}, X_2^{r'}, \ldots, X_{r'-1}^{r'}, X_1^{2r'}, X_2^{2r'}, \ldots, X_{r'-1}^{2r'}, \ldots, X_1^{r'(r'-1)}, X_2^{r'(r'-1)}, \ldots, X_{r'-1}^{r'(r'-1)}$. Given these values, the adversary can attack each of the r'-encryptions (e.g., $E^{[1\cdots r']}$), separately, and obtain 2^n "solutions" on average. Then, the adversary can treat each r'-round encryption as a single encryption with 2^n possible keys, and apply an r'-encryption attack to recover the key.

The time complexity of $Square_r$ is equivalent to repeating $2^{(r'-1)(r'-1)n}$ times a sequence of r' + 1 attacks on r'-encryption. Hence, the time complexity is at

most $2^{[(r'-1)(r'-1)+(r'-1)]\cdot n}$, and the memory complexity is kept at 2^n . Therefore, $Gain(Square_r) \geq \sqrt{r} - 1$, which shows that $Gain_D(r, 1) \geq |\sqrt{r}| - 1$.

Obviously, improving the time complexity of attacking r'-encryption with 2^n memory reduces the time complexity of $Square_r$ as well. However, as the best known attacks of this kind yields a gain of $O(\sqrt{r'}) = O(r^{1/4})$, the addition to the overall gain of $Square_r$ is negligible.

3.4 Asymmetric Dissections: 7-Encryption and Beyond

A common feature shared by the $LogLayer_r$ and the $Square_r$ algorithms is their symmetry. In both algorithms, every dissection partitions the composition into parts of the same size. In this section we show that a better gain can be achieved by an asymmetric dissection, and present the optimal dissection algorithms of this type for m=1 and any number r of encryptions.

We observe that the basic dissection attack is asymmetric in its nature. Indeed, after the two separate meet in the middle attacks are performed, the suggestions from the upper part are stored in a table, while the suggestions from the lower part are checked against the table values. As a result, the number of suggestions in the upper part is bounded from above by the size of the memory (which is now assumed to be 2^n and kept in sorted order), while the number of suggestions from the lower part can be arbitrarily large and generated on the fly in an arbitrary order. This suggests that an asymmetric dissection in which the lower part is bigger than the upper part, may result in a better algorithm. This is indeed the case, as illustrated by the following $Dissect_3(7,1)$ algorithm:

- 1. Given 7 plaintext-ciphertext pairs $(P_1, C_1), (P_2, C_2), \dots (P_7, C_7)$, for each possible value of X_1^3, X_2^3 , perform:
 - (a) Apply the basic meet in the middle algorithm to $E^{[1...3]}$ with (P_1, X_1^3) and (P_2, X_2^3) as the plaintext/ciphertext pairs, and obtain 2^n candidates for the keys (k_1, k_2, k_3) . For each such candidate, partially encrypt the rest of the plaintexts using (k_1, k_2, k_3) and store the values (X_4^3, \ldots, X_7^3) in a table, along with the corresponding key candidate (k_1, k_2, k_3) .
 - (b) Apply the algorithm $Dissect_2(4,1)$ to $E^{[4...7]}$ with (X_1^3, C_1) and (X_2^3, C_2) as the plaintext/ciphertext pairs. Note that since only two pairs are given, the algorithm $Dissect_2(4,1)$ produces 2^{2n} possible values of the keys (k_4, k_5, k_6, k_7) . However, these values are produced sequentially, and can be checked on-the-fly by partially decrypting C_4, C_5, C_6, C_7 , and checking whether the corresponding vector (X_4^3, \ldots, X_7^3) appears in the table

The memory complexity of the algorithm is 2^n , as both the basic meet in the middle attack on triple encryption and the algorithm $Dissect_2(4,1)$ require 2^n memory, and the size of the table "passed" between Steps 2(a) and 2(b) is also 2^n .

The time complexity is 2^{4n} . Indeed, two *n*-bit values are enumerated in the middle, both the basic meet in the middle attack on triple encryption and the

algorithm $Dissect_2(4,1)$ require 2^{2n} time, and the remaining 2^{2n} possible values of (k_4, k_5, k_6, k_7) are checked instantly. This leads to time complexity of $2^{2n} \cdot 2^{2n} = 2^{4n}$.

This shows that $Gain(Dissect_3(7,1)) = 2$, which is clearly better than the algorithms $LogLayer_r$ and $Square_r$, which reach gain of 2 for the first time only at r = 8 and at r = 9, respectively.

Furthermore, the algorithm $Dissect_3(7,1)$ can be extended recursively to larger values of r, to yield better asymptotic for the gain function. Given the algorithm $Dissect_j(r',1)$ such that $Gain(Dissect_j(r',1)) = \ell - 1$, we define the algorithm $Dissect_{NEXT}^1 = Dissect_{\ell+1}(r'+\ell+1,1)$ for r-encryption, where $r = r' + \ell + 1$, as follows:

- 1. Given r plaintext-ciphertext pairs $(P_1, C_1), (P_2, C_2), \dots (P_r, C_r)$, for each possible value of $X_1^{\ell+1}, \dots, X_\ell^{\ell+1}$, perform:
 - (a) Apply the basic meet in the middle algorithm to $E^{[1...\ell+1]}$ with $(P_1, X_1^{\ell+1}), \ldots, (P_\ell, X_\ell^{\ell+1})$ as the plaintext/ciphertext pairs, and obtain 2^n candidates for the keys $(k_1, \ldots, k_{\ell+1})$. For each such candidate, partially encrypt the rest of the plaintexts using $(k_1, \ldots, k_{\ell+1})$ and store the values $(X_{\ell+1}^{\ell+1}, \ldots, X_r^{\ell+1})$ in a table, along with the corresponding key candidate $(k_1, \ldots, k_\ell + 1)$.
 - a table, along with the corresponding key candidate $(k_1, \ldots, k_{\ell} + 1)$. (b) Apply the algorithm $Dissect_j(r', 1)$ to $E^{[\ell+2\ldots r]}$ with $(X_1^{\ell+1}, C_1), \ldots, (X_{\ell}^{\ell+1}, C_{\ell})$ as the plaintext/ciphertext pairs. Check each of the $2^{(r'-\ell)n}$ suggestions for the keys $(k_{\ell+2}, \ldots, k_r)$ on-the-fly by partially decrypting $C_{\ell+1}, \ldots, C_r$, and checking whether the corresponding vector $(X_{\ell+1}^{\ell+1}, \ldots, X_r^{\ell+1})$ appears in the table.

An exactly similar argument as the one used for $Dissect_3(7,1)$ shows that the time and memory complexities of $Dissect_{\ell+1}(r)$ are $2^{r'n}$ and 2^n , respectively, which implies that $Gain(Dissect_{\ell+1}(r)) = \ell$. In fact, $Dissect_3(7,1)$ can be obtained from $Dissect_2(4,1)$ by the recursive construction just described.

The recursion leads to a sequence of asymmetric dissection attacks with memory $M=2^n$, such that the gain increases by 1 with each step of the sequence. If we denote the number r of "rounds" in the ℓ 's element of the sequence (i.e., the element for which the gain equals to ℓ) by r_{ℓ} , then by the construction, the sequence satisfies the recursion

$$r_\ell = r_{\ell-1} + \ell + 1,$$

which (together with $r_2 = 4$ which follows from $Dissect_2(4,1)$) leads to the formula:

$$r_{\ell} = \frac{\ell(\ell+1)}{2} + 1.$$

The asymptotic gain of this sequence is obtained by representing ℓ as a function of r, and is equal to $(\sqrt{8r-7}-1)/2 \approx \sqrt{2r}$, which is bigger than the \sqrt{r} gain of the $Square_r$ algorithm.

A thorough analysis, presented in Appendix A.1, shows that the algorithms obtained by the recursive sequence described above are the optimal amongst all

dissection algorithms that split the r rounds into two (not necessarily equal) parts, and attacks each part recursively, using an optimal dissection algorithm.

We conclude that as far as only dissection attacks are concerned, the "magic sequence" of the minimal numbers of rounds for which the gains are $\ell = 0, 1, 2, 3, \ldots$, is:

$$Magic_1 = \{1, 2, 4, 7, 11, 16, 22, 29, 37, 46, 56, \ldots\}.$$

This "magic sequence" will appear several more times in the sequel.

4 Parallel Collision Search via Dissection

In Section 3, we considered the scenario of deterministic algorithms which never err for r-encryption, that is, algorithms which find all the possible values of the keys which comply with the plaintext/ciphertext pairs, or prove that there are no such keys. In this scenario, the best previously known generic attack is the meet in the middle attack, which obtains the tradeoff curve $TM = 2^{rn}$, where T and M are the time and memory complexities of the algorithm, respectively. In this model, we presented several dissection algorithms which allow to achieve the curve $TM = 2^{(r-\sqrt{2r})n}$.

In this section, we consider the scenario in which non-deterministic algorithms, which find the right keys with some probability p < 1, are allowed. In this case, an improved tradeoff curve of $T^2M = 2^{(3/2)rn}$ can be obtained by the parallel collision search algorithm of van Oorschot and Wiener [15]. We now show how to combine the dissection algorithms presented in Section 3 with the parallel collision search algorithm to obtain an even better tradeoff curve with a multiplicative gain of $2^{(\sqrt{2r}/8)n}$ over the curve of [15].

4.1 Brief Description of the Parallel Collision Search Algorithm

We start with a very brief description of the PCS algorithm suggested in [15]. For the sake of completeness, the full description of the algorithm and its analysis are given in Appendix C.1.

The algorithm consists of two steps:

- 1. Find partial collisions, which are key suggestions which comply with half of the plaintext/ciphertext pairs.
- 2. For each partial collision, check whether it complies with the second half of the plaintext/ciphertext pairs.

The first step is performed by constructing two step functions:

$$F^{upper}: (k_1, \dots, k_{r/2}) \mapsto (X_1^{r/2}, \dots, X_{r/2}^{r/2})$$
 and $F^{lower}: (k_{r/2+1}, \dots, k_r) \mapsto (X_1^{r/2}, \dots, X_{r/2}^{r/2}),$

which can be computed easily given the plaintext/ciphertext pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$, and using Floyd's cycle finding algorithm [8] to find a collision between them. In the case of constant memory, Floyd's algorithm finds such a collision (which produces a key suggestion which complies with the pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$)

in $2^{(r/4)n}$ time. If $M=2^{mn}$ memory is given, this step can be speeded up by incorporating Hellman's time-memory tradeoff techniques [4], that allow to find 2^{mn} collisions simultaneously in $2^{(r/4+m/2)n}$ time. In both cases, after $2^{(r/2)n}$ partial collisions are found, it is expected that one of them passes the condition of the second step, which means that it is the desired key suggestion. The time complexity of the algorithm is $T=2^{(r/4+m/2)n}\cdot 2^{(r/2-m)n}=2^{(3r/4-m/2)n}$, which leads to the tradeoff curve $T^2M=2^{(3/2)rn}$.

4.2 The Dissect & Collide Algorithm

In this section we present the Dissect & Collide (DC) algorithm, which uses dissection to enhance the PCS algorithm.

The basic idea behind the DC algorithm is that it is possible to fix several intermediate values after r/2 rounds, that is, $(X_1^{r/2}, \ldots, X_u^{r/2})$, and construct step functions \tilde{F}^{upper} and \tilde{F}^{lower} in such a way that all the keys they suggest partially encrypt P_i to $X_i^{r/2}$ and partially decrypt C_i to $X_i^{r/2}$, for all $i \leq u$. This is achieved by incorporating an attack on $E^{[1...r/2]}$ with $(P_1, X_1^{r/2}), \ldots, (P_u, X_u^{r/2})$ as the plaintext/ciphertext pairs into the function F^{upper} , and similarly with $E^{[r/2+1...r]}$ and F^{lower} . As a result, a partial collision which complies with the pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$ can be found at the smaller "cost" of finding a collision which complies only with $(P_{u+1}, C_{u+1}), \ldots, (P_{r/2}, C_{r/2})$. It should be noted that this gain could be diminished by the "cost" of the new step function \tilde{F} , that is higher than the "cost" of the simpler step function F. However, we show that if the efficient dissection algorithms presented in Section 3 are used to attack the subciphers $E^{[1...r/2]}$ and $E^{[r/2+1...r]}$, the gain is bigger than the loss, and the resulting DC algorithm is faster than the PCS algorithm (given the same amount of memory).

A basic example: Applying DC to 8-encryption As the idea of the DC algorithm is somewhat involved, we illustrate it by considering the simple case (r = 8, m = 1). In the case of 8-encryption, the goal of the first step in the PCS algorithm is to find partial collisions which comply with the pairs $(P_1, C_1), \ldots, (P_4, C_4)$. Given memory of 2^n , the average time PCS requires for finding each such collision is $2^{1.5n}$. The DC algorithm allows to achieve the same goal in 2^n time.

In the DC algorithm, we fix three intermediate values: (X_1^4, X_2^4, X_3^4) , and want to attack the subciphers $E^{[1...4]}$ and $E^{[5...8]}$. Recall that the algorithm $Dissect_2(4,1)$ presented in Section 3 allows to retrieve all 2^n values of (k_1,k_2,k_3,k_4) which comply with the pairs $(P_1,X_1^4),(P_2,X_2^4),(P_3,X_3^4)$ in time 2^{2n} and memory 2^n . Furthermore, given a fixed value X_1^2 , there is a single value of (k_1,k_2,k_3,k_4) (on average) which complies with the three plaintext/ciphertext pairs and the X_1^2 value, and this value can be found in time 2^n (since the $Dissect_2(4,1)$ algorithm starts with guessing the value X_1^2 and then performs only 2^n operations for each guess).

The algorithm works as follows:

- 1. Given plaintexts (P_1, P_2, P_3, P_4) and their corresponding ciphertexts (C_1, C_2, C_3, C_4) , for each guess of (X_1^4, X_2^4, X_3^4) : 2. (a) Define the step functions \tilde{F}^{upper} and \tilde{F}^{lower} by:

$$\tilde{F}^{upper}: X_1^2 \mapsto X_4^4$$
 and $\tilde{F}^{lower}: X_1^6 \mapsto X_4^4$.

In order to compute the step function \tilde{F}^{upper} , apply the $Dissect_2(4,1)$ algorithm to $E^{[1...4]}$ with the plaintext/ciphertext pairs $(P_1, X_1^4), (P_2, X_2^4), (P_3, X_3^4)$ and the intermediate value X_1^2 to obtain a unique value of the keys (k_1, k_2, k_3, k_4) . Then, partially encrypt P_4 through $E^{[1...4]}$ with these keys to obtain $\tilde{F}^{upper}(X_2^1) = X_4^4$. The function \tilde{F}^{lower} is computed similarly.

- (b) Find a collision between the functions \tilde{F}^{upper} and \tilde{F}^{lower} using a variant of Floyd's cycle finding algorithm which exploits the $M=2^n$ available amount of memory.
- (c) Check whether the keys $(k_1, \ldots, k_4, k_5, \ldots, k_8)$ suggested by the partial collision, encrypt (P_5, \ldots, P_8) to (C_5, \ldots, C_8) . If not, return to Step 2(a). After 2^n partial collisions are examined and discarded, return to Step 1.

By the properties of the algorithm $Dissect_2(4,1)$ mentioned above, each step of the functions \tilde{F} can be performed in 2^n time and memory. By the construction of the step functions, each suggested key (k_1, \ldots, k_4) (or (k_5, \ldots, k_8)) encrypts (P_1, P_2, P_3) to (X_1^4, X_2^4, X_3^4) (or decrypts (k_5, \ldots, k_8) to (X_1^4, X_2^4, X_3^4) , respectively), and hence, each collision between \tilde{F}^{upper} and \tilde{F}^{lower} yields a suggestion of $(k_1,\ldots,k_4,k_5,\ldots,k_8)$ which complies with the pairs $(P_1,C_1),\ldots,(P_4,C_4)$. Finally, since the step functions are from n bits to n bits, collision between them can be found instantly given 2^n memory. Therefore, the time required for finding a partial collision is 2^n , and thus, the total running time of the algorithm is $2^{4n} \cdot 2^n = 2^{5n}$. We note that while our DC algorithm outperforms the respective PCS algorithm (whose time complexity is $2^{5.5n}$), it has the same performance as the $Dissect_4(8,1)$ algorithm presented in Section 3. However, as we will show in the sequel, for larger values of r, the DC algorithms outperform the Dissectalgorithms significantly.

The general algorithms DC(r, m) Now we are ready to give a formal definition of the class DC(r, m) of algorithms, applicable to r-encryption (for an even $r)^4$, given memory of 2^{mn} . An algorithm $A \in DC(r,m)$ is specified by a number $u, 1 \leq u \leq r/2$, and two sets I^{upper} and I^{lower} of intermediate locations in the subciphers $E^{[1...r/2]}$ and $E^{[r/2+1...r]}$, respectively, such that $|I^{upper}| = |I^{lower}| = r/2 - u.$

In the algorithm, the adversary fixes u intermediate values $(X_1^{r/2}, \dots, X_u^{r/2})$. Then, she defines the step functions \tilde{F}^{upper} and \tilde{F}^{lower} by:

$$\tilde{F}^{upper}: I^{upper} \mapsto (X^{r/2}_{u+1}, \dots, X^{r/2}_{r/2}) \quad \text{ and } \quad \tilde{F}^{lower}: I^{lower} \mapsto (X^{r/2}_{u+1}, \dots, X^{r/2}_{r/2}).$$

⁴ We note that for sake of simplicity, we discuss in this section only even values of r. An easy (but probably non-optimal) way to use these algorithms for an odd value of r is to guess the value of the key k_r , and for each guess, to apply the algorithms described in this section to $E^{[1...r-1]}$.

The step function \tilde{F}^{upper} is computed by applying a dissection attack to $E^{[1...r/2]}$ with the plaintext/ciphertext pairs $(P_1, X_1^{r/2}), \ldots, (P_u, X_u^{r/2})$ and the intermediate values contained in I^{upper} to retrieve a unique value of the keys $(k_1, \ldots, k_{r/2})$, and then partially encrypting (P_{u+1}, \ldots, P_r) with these keys to obtain $(X_{u+1}^{r/2}, \ldots, X_{r/2}^{r/2})$. The step function \tilde{F}^{lower} is computed in a similar way, with respect to $E^{[r/2+1...r]}$ and the set I^{lower} . Then, a variant of Floyd's cycle finding algorithm which exploits the 2^{mn} amount of available memory is used to find a collision between \tilde{F}^{upper} and \tilde{F}^{lower} , which yields a suggestion of $(k_1, \ldots, k_{r/2}, k_{r/2+1}, \ldots, k_r)$ which complies with the plaintext/ciphertext pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$.

Denote the time complexity of each application of \tilde{F} by $S=2^{sn}$. An easy computation shows that the overall time complexity of the algorithm DC(r,m) is:

$$2^{(r/2)n} \cdot 2^{((r/2-u-m)/2)n} \cdot 2^{sn} = 2^{((3/4)r - (u+m-2s)/2)n}.$$
 (1)

As the time complexity of the PCS algorithm with memory 2^{mn} is $2^{((3/4)r-m/2)n}$, the multiplicative gain of the DC algorithm is $2^{(u/2-s)n}$. In particular, for the specific DC(8,1) algorithm described above for 8-encryption, we have s=1, and thus, the gain is indeed $2^{(3/2-1)n}=2^{n/2}$, as mentioned above. In the sequel, we denote the parameters $I^{upper}, I^{lower}, u, s$ which specify a DC(r, m) algorithm A and determine its time complexity by $I^{upper}(A), I^{lower}(A), u(A)$, and s(A), respectively.

We conclude this section by mentioning a difficulty in the implementation of the DC algorithm. Unlike the PCS algorithm where the output of the step functions F is always uniquely defined, in DC the functions \tilde{F} return no output for some of the inputs. This happens since the number of keys $(k_1, \ldots, k_{r/2})$ which comply with the u plaintext/ciphertext values $(P_1, X_1^{r/2}), \ldots, (P_u, X_u^{r/2})$ and the r/2 - u fixed intermediate values contained in I^{upper} , is distributed according to the distribution Poisson(1), and in particular, equals to zero for an 1/e fraction of the inputs. This difficulty can be resolved by introducing flavors into the step function \tilde{F} , which alter the function in a deterministic way when it fails to produce output. The exact modification is described in Appendix C.2.

4.3 The Gain of the Dissect & Collide Algorithm Over the PCS Algorithm

In this section we consider several natural extensions of the basic DC(8,1) algorithm presented in Section 4.2. We use these extensions to show that the gain of the DC algorithms over the PCS algorithm is monotone non-decreasing with r and is lower bounded by $2^{(\lfloor \sqrt{2r}\rfloor/8)n}$ for any $r \geq 8$.

Before we present the extensions of the basic DC algorithm, we would like to define formally the notion of gain in the non-deterministic setting. As the best previously known algorithm in this setting is the PCS algorithm, whose time complexity given 2^{mn} memory is $2^{((3/4)r-m/2)n}$, we define the gain with respect to it.

Definition 2. The gain of a probabilistic algorithm A for r-encryption whose time and memory complexities are T and $M = 2^{mn}$, respectively, is defined as

$$Gain_{ND}(A) = (3/4)r - m/2 - (\log T)/n.$$

The maximal gain amongst all probabilistic DC algorithms for r-encryption which require 2^{mn} memory, is denoted by $Gain_{ND}(r,m)$.

Note that it follows from Equation (1) that if $A \in DC(r, m)$, then

$$Gain_{ND}(A) = u(A)/2 - s(A).$$
(2)

Monotonicity of the gain The most basic extension of the basic DC algorithm is to preserve the gain when additional "rounds" are added. While in the deterministic case, such an extension can be obtained trivially by guessing several keys and applying the previous algorithm, in our setting this approach leads to a decrease of the gain by 1/2 for each two added rounds (as the complexity of the PCS algorithm is increased by a factor of $2^{3n/2}$ when r is increased by 2). However, the gain can be preserved in another way, as shown in the following lemma.

Lemma 1. Assume that an algorithm $A \in DC(r', m)$ has gain ℓ . Then there exists an algorithm $B \in DC(r'+2, m)$ whose gain is also equal to ℓ .

Due to space restrictions, the proof of the lemma is presented in Appendix C.3. Here we only note that the algorithm B is constructed from A by choosing $I^{upper}(B) = I^{upper}(A) \cup \{X_1^{r'/2}\}$, and similarly for $I^{lower}(B)$.

Lemma 1 implies that the gain of the DC algorithms is monotone non-decreasing with r, and in particular, that $Gain_{ND}(r,1) \geq 1/2$, for any even $r \geq 8$.

An analogue of the LogLayer algorithm The next natural extension of the basic DC algorithm is an analogue of the LogLayer algorithm presented in Section 3.3. Recall that the $LogLayer_r$ algorithm, applicable when r is a power of 2, consists of guessing the set of intermediate values:

$$I_0 = \{X_1^2, X_1^4, \dots, X_1^{r-2}, X_2^4, X_2^8, \dots, X_2^{r-4}, X_3^8, \dots, X_3^{r-8}, \dots, X_{\log r-1}^{r/2}\},\$$

and applying a recursive sequence of meet in the middle attacks on 2-encryption. Using this algorithm, we can define the algorithm $LL_r \in DC(2r, 1)$, by specifying $I^{upper}(LL_r) = I_0$, and $I^{lower}(LL_r)$ in a similar way. Since $|I_0| = r - \log r - 1$, we have $u(LL_r) = r - (r - \log r - 1) = \log r + 1$. It follows from the structure of the $LogLayer_r$ algorithm that given the values in I_0 , it can compute the keys (k_1, \ldots, k_r) in time and memory of 2^n . Hence, we have $s(LL_r) = 1$. By Equation (2), it follows that $Gain(LL_r) = (\log r + 1)/2 - 1 = (\log r - 1)/2$.

The basic algorithm for 8-encryption is the special case LL_4 of this algorithm. The next two values of r also yield interesting algorithms: LL_8 yields gain of 1 for (r = 16, m = 1), which amounts to an attack on 16-encryption with $(T = 2^{10.5n}, M = 2^n)$, and LL_{16} yields gain of 1.5 for (r = 32, m = 1), which amounts to an attack on 32-encryption with $(T = 2^{22n}, M = 2^n)$. Both attacks outperform the *Dissect* attacks and are the best known attacks on 16-encryption and on 32-encryption, respectively.

An analogue of the $Square_r$ algorithm: The logarithmic asymptotic gain of the LL sequence can be significantly outperformed by an analogue of the $Square_r$ algorithm, presented in Section 3.3. Recall that the $Square_r$ algorithm, applicable when $r = (r')^2$ is a perfect square, starts by guessing the set of $(r'-1)^2$ intermediate encryption values:

$$I_1 = \{X_1^{r'}, X_2^{r'}, \dots, X_{r'-1}^{r'}, X_1^{2r'}, X_2^{2r'}, \dots, X_{r'-1}^{2r'}, \dots, X_1^{r'(r'-1)}, X_2^{r'(r'-1)}, \dots, X_{r'-1}^{r'(r'-1)}\},\$$

and then performs a two-layer attack, which amounts to r'+1 separate attacks on r'-encryption. Using this algorithm, we can define the algorithm $Sq_r \in DC(2r,1)$, by specifying $I^{upper}(Sq_r) = I_0$, and $I^{lower}(Sq_r)$ in a similar way. Since $|I_0| = (r'-1)^2$, we have $u(Sq_r) = r - (r'-1)^2 = 2r' - 1$. The step complexity $s(Sq_r)$ is the time complexity required for attacking r'-encryption without fixed intermediate values. Hence, by Equation (2),

$$Gain(Sq_r) = r' - 1/2 - f_1(r'),$$

where $2^{f_1(r)n}$ is the time complexity of the best possible attack on r-encryption with 2^n memory.

The basic algorithm for 8-encryption is the special case Sq_2 of this algorithm. Since for small values of r', the best known attacks on r'-encryption are obtained by the dissection attacks presented in Section 3.4, the next elements of the sequence Sq_r which increase the gain, correspond to the next elements of the sequence $Magic_1 = \{1, 2, 4, 7, 11, 16, \ldots\}$ described in Section 3.4. They lead to gains of 1.5, 2.5, and 3.5 for r = 32, 98, and r = 242, respectively. For large values of r, the PCS algorithm outperforms the Dissect algorithms, and using it we obtain:

$$Gain(Sq_r) > r' - 1/2 - ((3/4)r' - 1/2) = r'/4 = \sqrt{2r}/8.$$

This shows that the asymptotic gain of the DC algorithms is at least $\sqrt{2r}/8$.

We note that as for $r' \geq 16$, the DC algorithm outperforms both the Dissect and the PCS algorithms, we can use it instead of PCS in the attacks on r'-encryption in order to increase the gain for large values of r. However, as the gain of DC over PCS for r'-encryption is only of order $O(\sqrt{r'}) = O(r^{1/4})$, the addition to the overall gain of Sq_r is negligible.

Two-layer DC algorithms A natural extension of the Sq_r algorithm is the class of two-layer DC algorithms. Assume that $r = 2r_1 \cdot r_2$, and that there exist algorithms A_1, A_2 for r_1 -encryption and for r_2 -encryption, respectively, which

perform in time 2^{sn} and memory 2^n given sets of intermediate values I_1^{upper} and I_2^{upper} , respectively.

Then we can define an algorithm $A \in DC(r,1)$ whose step function is computed by a two-layer algorithm: First, $E^{[1\dots r/2]}$ is divided into r_2 subciphers of r_1 rounds each, and the algorithm A_1 is used to attack each of them separately and compute 2^n possible suggestions for each set of r_1 consecutive keys. Then, each r_1 -round encryption is considered as a single encryption with 2^n possible keys, and the algorithm A_2 is used to attack the resulting r_2 -encryption. The set $I^{upper}(A)$ is chosen such that both A_1 and A_2 algorithms perform in time 2^s . Formally, if we denote $u_1 = |I_1^{upper}|$, then the set $I^{upper}(A)$ consists of r_2 "copies" of the set I_1^{upper} , $r_1 - 1 - u_1$ intermediate values after each r_1 rounds, and one copy of the set I_2^{upper} . The set $I^{lower}(A)$ is defined similarly. Hence,

$$u(A) = r/2 - |I^{upper}(A)| = r/2 - (r_2 \cdot u_1 + (r_2 - 1)(r_1 - 1 - u_1) + u_2) = r_2 + r_1 - u_1 - u_2 - 1.$$

As
$$s(A) = s$$
, we have $Gain_{ND}(A) = (r_2 + r_1 - u_1 - u_2 - 1)/2 - s$.

Note that the algorithm Sq_r is actually a two-layer DC algorithm, with $r_1=r_2=r'$ and $I_1^{upper}=I_2^{upper}=\emptyset$. It turns out that for all $8 \le r \le 128$, the maximal gains are obtained by two-layer DC algorithms where r_1, r_2 are chosen from the sequence $Magic_1$ presented in Section 3.4, and A_1, A_2 are the respective Dissect algorithms. The cases of r=8,16,32 presented above are obtained with $r_1=4$ and $r_2=1,2,4$ (respectively), and the next numbers of rounds in which the gain increases are r=56,88,128, obtained for $r_1=4$ and $r_2=7,11,16$, respectively. The continuation of the "non-deterministic magic sequence" is, however, more complicated. For example, the two-layer algorithm for r=176 with $(r_1=4,r_2=22)$ has the same gain as the algorithm with $(r_1=4,r_2=16)$, and the next increase of the gain occurs only for r=224, and is obtained by a two-layer algorithm with $(r_1=7,r_2=16)$. For larger values of r, more complex algorithms, such as a three-layer algorithms. We leave the analysis of the whole magic sequence to the full version of the paper, and conclude that the minimal numbers of rounds for which the gain equals $0.5, 1, 1.5, \ldots$ are:

$$Magic_1^{ND} = \{8, 16, 32, 56, 88, 128, \ldots\}.$$

Finally, we note that two-layer DC algorithms can be applied also for m > 1, and can be used to show that the first numbers of rounds for which $Gain_{ND}(r, m) = 0.5, 1, 1.5, 2, \dots$ are:

$$Magic_m^{ND} = \{8m, 8m + 8, 8m + 16, \dots, 16m, 16m + 16, 16m + 32, \dots, 32m, 32m + 24, 32m + 48, \dots, 56m, \dots\}.$$

The full analysis of the case m > 1 will appear in the full version of the paper.

5 Applications

In this section, we apply our new dissection algorithms to several well known bicomposite search problems. As described in the introduction, we can represent

such a problem as an $r \times r$ execution matrix which is treated as a multiple-encryption scheme with r rounds. In the case of knapsacks, we are allowed to choose any constant value of r when n grows to infinity in order to optimize the value of t for any given m. In other cases, r is restricted to a specific value or to a set of values. For example, in the special case of Rubik's cube, we know that a 20-move solution exists for any state, and thus it does not make sense to choose r > 20. Such constraints can limit the choice of parameters for our algorithms, and thus we may not be able to exploit the available memory as efficiently as in the case of knapsacks.

Since the analysis of our multiple encryption algorithms assumes the randomness of the underlying block ciphers, we have to justify this assumption for each reduction we consider. For example, in the case of knapsacks with n generators, strong randomness assumptions are common practice whenever n is sufficiently large (e.g., see [1,6]), and we can use the same assumptions when we consider subproblems with n/r generators for any constant r.

5.1 Applications to Knapsacks

The knapsack problem is a well-known problem that has been studied for many years. For more than 30 years, the best known algorithm for knapsacks was the Schroeppel-Shamir algorithm [14], which requires $2^{n/2}$ time and $2^{n/4}$ memory. Surprisingly, in 2010, Howgrave-Graham and Joux [6] showed how to solve the knapsack problem in time much better than $2^{n/2}$, by using the associativity and commutativity properties of addition. This result was further improved by Becker, Coron and Joux in [1]. In addition to their basic results, Howgrave-Graham and Joux [6] also described reduced-memory algorithms, and in particular [1] described a memoryless attack which requires only $2^{0.72n}$ time. All these new attacks are heuristic in a sense that they may fail to find a solution even when it exists, and thus they cannot be used in order to prove the nonexistence of solutions. In addition to these heuristic algorithms, Becker, Coron and Joux [1] also considered deterministic algorithms that never err, and described a straight-line time-memory tradeoff curve, but this curve was only valid in the range $1/16 \le m \le 1/4$.

In this section, we show how to use our generic dissection techniques in order to find deterministic algorithms for the knapsack problem which are better than the deterministic tradeoff curve described in [1] over the whole range of 1/16 < m < 1/4. In addition, we can expand our tradeoff curve in a continuous way for any smaller value of $m \le 1/4$. By combining our generic deterministic and non-deterministic algorithms, we can get a new curve which is better than the best knapsack-specific algorithms described in [6] and [1] for the large interval of (approximately) $1/100 \le m < 1/6$.

The formal reduction of the knapsack problem to r-round encryption (for any r) is given in Appendix B, but it is not required in order to understand the rest of this paper. Given $M=2^{mn}$ memory, our goal is to solve the knapsack problem by applying the reduction with a value of r which optimizes the time complexity of our multiple encryption algorithm. Formally, for any r, we apply

the multiple encryption algorithm with an effective block size reduced by a factor of r, i.e., $n^* = n/r$. By equating $M = 2^{mn}$ with $M = 2^{n(m^*r)}$, we can see that the effective memory unit increases by the same ratio, i.e, $m^* = mr$. We denote by $f(r, n^*, m^*)$ the running time of our multiple encryption algorithm on an r-round block cipher with a block size of n^* bits and $M^* = 2^{m^*n^*}$ available memory, given r plaintext-ciphertext pairs. Using this notation, we would like to find r that minimizes $f(r, n^*, m^*) = f(r, n/r, mr)$. We call such a value of r an optimal value.

We note that the deterministic algorithms applied in [6] and [1] for $1/16 \le m \le 1/4$ implicitly perform a reduction to multiple encryption with the fixed parameters r=4 and r=16. In fact, for the case of knapsacks and these choices of r, these algorithms are closely related to our square algorithms (described in Section 3.3). However, as we now show, we can get a better tradeoff curve by using other choices of r.

Time-Memory Tradeoff Curves for Knapsacks Using our multiple encryption algorithms, we construct time-memory tradeoff curves for knapsacks: we start with deterministic algorithms and consider first the case of $1/m \in$ $\{1, 2, 4, 7, 11, 16, \ldots\}$, which is the "magic sequence" constructed in Section 3.4.⁵ In order to simplify our notation, we denote the j'th element of this sequence by b_j , starting from j=0. In the case of $1/m=b_j$ for $j\geq 2$, we simply choose $r=1/m=b_j$ and run the algorithm with $n^*=n/m$ and $m^*=m/m=1$. For example, in case m = 1/4, we run the 4-round multiple encryption with $m^* = 1$, for which the time complexity is $T = 2^{2n^*} = 2^{2(n/4)} = 2^{n/2}$, or t = 1/2. In case m=1/7, we run the 7-round dissection algorithm with $m^*=1$, for which the time complexity is $T = 2^{4n^*} = 2^{4n/7}$, or t = 4/7. In Appendix B, we show that in the case of $1/m \in \{4, 7, 11, 16, ...\}$, our choice of r is indeed optimal. Thus, we obtain a sequence of optimal points on the deterministic time-memory tradeoff curve. In order to obtain an optimal continuous curve for $0 < m \le 1/4$, we need to use our algorithms for integral $m^* \geq 2$. As described in Appendix B, these algorithms enable us to connect in a straight line any two consecutive time-memory tradeoff points for $1/m \in \{4, 7, 11, 16, ...\}$, and obtain a continuous curve (as shown on the left diagram of Figure 1).

For non-deterministic algorithms, we use the same approach, and consider first the "magic sequence" constructed in Section 4 for $1/m \in \{16, 32, 56, ...\}$. We choose r = 1/m and the corresponding values of t (1/10.5, 1/22, 1/39.5,...). Similarly to the deterministic case, we can use our non-deterministic algorithms for integral $m^* \geq 2$ in order to obtain a continuous curve (as shown on the right diagram of Figure 1). The full details of how to connect consecutive points on the curve will be given in the full version of this paper.

⁵ Since our algorithms do not need the additional memory of m > 1/4, we actually consider only m < 1/4.

5.2 Improving Rebound Attacks On Hash Functions

Another application of our new techniques can significantly improve rebound attacks [11] on hash functions. An important procedure in such attacks is to match input/output differences through an S-box layer (or a generalized S-box layer). More precisely, the adversary is given a list L_A of input differences and a list L_B of output differences, and has to find all the input/output difference pairs that can happen through the S-box layer. A series of matching algorithms were recently developed, optimizing and improving various rebound attacks [13].

Our dissection algorithms can be applied for this problem as well, replacing the gradual matching or parallel matching presented at Crypto 2011 by [13]. For example, we can improve the rebound attack on Luffa using a variant of our $Dissect_2(4,1)$ algorithm. As described in Appendix A.5, we can reduce the memory complexity of the matching algorithm from 2^{102} to only 2^{66} without affecting the time complexity of the attack (which remains at 2^{104}).

5.3 Applications to Relational Databases

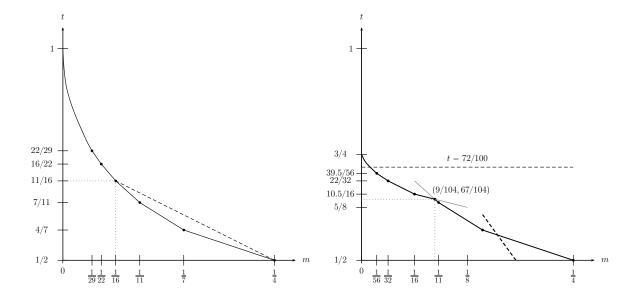
As a final example of the versatility of our algorithm, we note that in some cases, it may be possible to use the dissection technique to speed up the processing of queries in relational databases. The problem of composing block ciphers can be viewed as the problem of computing the *join* of several databases, where each database contains all the possible plaintext/ciphertext pairs, and the join operation equates the previous ciphertext with the next plaintext. When intermediate joined databases blow up in size but the final database is quite small, it may be better to use the dissection technique which guesses some middle values and splits the computation into smaller independent parts. More details about this potential application will be given in the full version of this paper.

6 Summary and Open Problems

In this paper we introduced the new dissection technique which can be applied to a broad class of problems which have a bicomposite structure. We used this technique to obtain improved complexities for several well studied problems such as the cryptanalysis of multiple-encryption schemes and the solution of hard knapsacks. The main open problem in this area is to either improve our techniques or to prove their optimality. In particular, we conjecture (but can not prove) that any attack on multiple-encryption schemes should have a time complexity which is at least the square root of the total number of possible keys.

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On the left: A comparison between our time-memory tradeoff curve and the curve obtained in [1] (shown as a dashed line) for deterministic algorithms. Our curve (defined for $m \le 1/4$) is strictly better than the curve obtained in [1] (defined only for $1/16 \le m \le 1/4$) for any 1/16 < m < 1/4.

On the right: A comparison between our general time-memory tradeoff curve, the curve obtained by extending the (m=0.211,t=0.421) attack given in [6] (shown as a bold dashed line), and the memoryless attack with t=0.72 obtained in [1] (shown as a light dashed horizontal line). Our general time-memory tradeoff curve is better than the attacks of [1] and [6] in the interval of (approximately) $1/100 \le m < 1/6$.

Fig. 1. Time-Memory Tradeoff Curves for Knapsack

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A Further Analysis of Deterministic Dissection Algorithms

In this appendix we present more detailed analysis of deterministic dissection algorithms for multiple encryption. First we present a thorough analysis of the algorithms $Dissect_u(r,1)$ and $Dissect_u(r,m)$, and show that they lead to optimal gain among the deterministic dissection algorithms. We show how to adapt the algorithms to the case when encryptions are replaced by one-way functions, and finally we discuss its application to the rebound attacks.

A.1 The $Dissect_u(r, 1)$ Algorithm

Consider the most generic form of a dissection algorithm for r-encryption. In the outer loop of the algorithm, the adversary dissects $E^{[1...r]}$ into two parts, $E^{[1...u]}$ and $E^{[u+1...r]}$, and guesses a few of the X_i^u values. Then she finds candidates for the keys k_1, k_2, \ldots, k_u by attacking $E^{[1...r]}$, and stores in a table their values, along with some additional X_j^u values. At this point, the adversary attacks $E^{[u+1...r]}$, deduces the candidate values for $k_{u+1}, k_{u+2}, \ldots, k_r$, computes

the corresponding X_j^u values, and looks for collisions in the table (each suggesting a value for the entire key of $E^{[1...r]}$). Obviously, the attacks on $E^{[1...u]}$ and $E^{[u+1...r]}$ themselves, can be done using dissection algorithms.

We note that similarly to the classical MITM attack, the partial keys that are obtained from $E^{[u+1\dots r]}$ do not have to be stored in memory. Instead, we verify them "on the fly" by using each one to partially decrypt additional ciphertexts, and checking for matches with the additional X^u_j values from the top part of the cipher.

The main question for this sort of an attack, is the amount of X_i^u values that are needed for the attack to succeed. As we limit the amount of memory to 2^n , then the size of the table constructed in the attack on $E^{[1...r]}$ cannot be larger than that (and the same goes for the two attacks on $E^{[1...u]}$ and $E^{[u+1...r]}$). The size of the external table is actually dictated by the number of possible keys returned by the attack on $E^{[1...u]}$. Assuming standard randomness assumptions on E itself, one can easily see that given u-1 "ciphertexts" (i.e., X_i^u values) we expect that the number of candidate keys k_1, k_2, \ldots, k_u is indeed 2^n .

The $Dissect_u(r, 1)$ Algorithm: We now present the generic $Dissect_u(r, 1)$ attack on r-encryption:

- 1. Given r plaintext-ciphertext pairs (P_1, C_1) , (P_2, C_2) , ... (P_r, C_r) , for each possible value of $X_1^u, X_2^u, \ldots, X_{u-1}^u$ perform:
 - (a) Obtain the 2^n candidates keys for k_1, k_2, \ldots, k_u given the plaintext"ciphertext" pairs $(P_1, X_1^u), (P_2, X_2^u), \ldots, (P_{u-1}, X_{u-1}^u)$. For each such candidate key compute $E^{[1...u]}(P_u, P_{u+1}, \ldots)$ and store the outcome in a table (along with the corresponding key candidate k_1, k_2, \ldots, k_u).
 - (b) Obtain the $2^{[r-u-(u-1)]n}$ candidate keys⁶ $k_{u+1}, k_{u+2}, \ldots, k_r$, and compute $D^{[u+1...r]}(C_u, C_{u+1}, \ldots)$ for each such candidate and check whether the corresponding value is in the table. If so, check the key candidate offered by the value of the current candidate of $k_{u+1}, k_{u+2}, \ldots, k_r$ with the value stored in the table for k_1, k_2, \ldots, k_u .

Complexity Analysis of $Dissect_u(r, 1)$: It is easy to see that if the memory complexity of the attacks on $E^{[1...u]}$ and $E^{[u+1...r]}$ is indeed 2^n , then so does the memory complexity of the whole attack (recall that we expect 2^n candidates for $k_1, k_2, ..., k_u$). Moreover, for the right guess of the $X_1^u, X_2^u, ... X_{u-1}^t$ values, we are assured that the right key value is suggested by both Step 2(a) and Step 2(b), and that for the right values, there is a collision in the table with respect to the other X_j^u values. Hence, we conclude that the attack indeed returns the right key (maybe with a few other candidates).

The remaining issue with respect to this attack algorithm is the expected running time. The running time is $2^{(u-1)n}$ times the repetition of Steps 2(a) and

⁶ We note that when attacking $E^{[u+1...n]}$, we expect that out of the $2^{(r-u)n}$ possible keys, only one in $2^{-(u-1)n}$ is consistent with the given given u-1 "plaintext"-ciphertext pairs.

2(b). The running time of Step 2(a) is equal to at most to 2^{un} (if implemented by a simple exhaustive search), but can be improved by using a more optimized algorithm (e.g., $Dissect_{u'}(u, 1)$ for some u' < u).⁷

The running time of Step 2(b) is equal to the time complexity of attacking (r-u)-encryption, which can be done either by exhaustive search, or by calling the $Dissect_{u^*}(r-u,1)$ algorithm for some $u^* < r-u$. However, one needs to note that the number of solutions suggested by this part of the attack is $2^{(r-2u+1)n}$ solutions, i.e., we expect another $2^{(r-2u+1)n}$ accesses to the table as part of Step 2(b).

For convenience of notation, we define by $f_1(r)$ the exponent divided by n of the time complexity of the optimal attack⁸ on r-encryption with 2^n memory. In other words, the optimal attack on r-encryption with 2^n memory takes $2^{f_1(r)n}$ time. Note that $f_1(r) = r - 1 - Gain_D(r, 1)$.

We recall that following the standard meet in the middle attacks, $f_1(2) = 1$ and $f_1(3) = 2$, and following Section 3.2 we obtain that $f_1(4) = 2$ as well. Hence, the time complexity of $Dissect_u(r,1)$ is equal to that of $2^{(u-1)n}$ repetitions of $Dissect_{u'}(u,1)$ and $Dissect_{u^*}(r-u,1)$ plus $2^{(r-2u+1)n}$ accesses to a table. Obviously, we suggest picking u' and u^* optimally, and thus, the time complexity of $Dissect_u(r,1)$ is

$$2^{(u-1)n} \cdot \max \left\{ 2^{f_1(u)n}, 2^{f_1(r-u)n}, 2^{(r-2u+1)n} \right\}.$$

Therefore, for a given value of r, the optimal value of u is the one that minimizes the above expression. In other words:

$$f_1(r) = \min_{1 \le u \le r} \left\{ u - 1 + \max \left\{ f_1(u), f_1(r - u), r - 2u + 1 \right\} \right\}.$$
 (3)

Using Equation 3 and the known values of $f_1(1) = f_1(2) = 1$, it is easy to show by induction on r that the minimal complexities are achieved by the sequence of algorithms presented in Section 3.4.

Remark 1. One may be concerned by the fact that when the algorithm is run recursively there are no additional "plaintext" values that allow constructing the table in Step 2(a) (as we require at least one additional "plaintext" to filter some of the key candidates found in Step 2(b)). This issue is solved by the fact that once the top part (being $E^{[1...r^*]}$ for some r^*) has at most 2^n possible keys, we can partially encrypt as many plaintexts as we have, to generate the required data.

A.2 Optimal Deterministic Dissection Algorithms for m > 1

While the algorithms presented above seem tailored to the case m = 1, it turns out that a small tweak in the recursive sequence of dissection algorithms presented in Section 3.4 is sufficient to obtain optimal algorithms for any m > 1.

⁷ As there are 2^n solutions for the keys, the generation of the table takes an additional 2^n time, but as any attack on the first u encryptions takes at least the same amount of time, it has no effect on the total running time of the attack.

 $^{^{8}}$ Optimal in that sense means lowest time complexity.

We start with the recursion step. Assume that we are given an algorithm $Dissect_j(r',m)$ for r'-encryption with memory of 2^{mn} , such that $Gain(Dissect_j(r',m)) = \ell-m$. We can define the algorithm $Dissect_{NEXT}^m = Dissect_{\ell+m}(r'+\ell+m,m)$ for r-encryption, where $r = r' + \ell + m$, similarly to the definition of $Dissect_{NEXT}^1$. Concretely, the adversary enumerates ℓ values at the dissection point, which is now placed after $\ell + m$ rounds. Then she performs a simple meet in the middle attack on the upper $\ell + m$ rounds and stores the 2^{mn} key suggestions for the upper part in a table. Then, the adversary applies the algorithm $Dissect_j(r',m)$ to the bottom part, and checks the $2^{(r'-\ell)n}$ suggested subkeys on-the-fly against the table. Exactly the same analysis as in the case of $Dissect_{NEXT}^1$ shows that the time and memory complexities of $Dissect_{\ell+m}(r' + \ell + m, m)$ are $2^{r'n}$ and 2^{mn} , respectively, and hence, $Gain(Dissect_{\ell+m}(r' + \ell + m, m)) = \ell$.

In this case, the "jump" of the gain between two consecutive elements of the sequence is m > 1, and hence, we need m starting points in order to complete the sequence. These starting points are given by the m algorithms: $Dissect_{m+i}(2m+2i,m)$, which have gains of $0,1,2,\ldots,m-1$ for $i=0,1,\ldots,m-1$, respectively. In the algorithm $Dissect_{m+i}(2m+2i,m)$, the adversary enumerates i intermediate values after m+i rounds, applies simple meet in the middle attacks on each part separately, and then applies a meet in the middle attack between the 2^{mn} key suggestions from the two parts. The time and memory complexities of the algorithm are $2^{(m+i)n}$ and 2^{mn} , respectively, and hence, its gain is indeed i.

Using these m starting points, the entire sequence can be computed easily. It turns out that the "magic sequence" of the minimal numbers of rounds for which the gain $\ell=0,1,2,\ldots$ is obtained is

$$Magic_m = \{2m, 2m+2, 2m+4, \dots, 4m, 4m+3, 4m+6, \dots, 7m, 7m+4, 7m+8, \dots, 11m, \dots\},\$$

and the asymptotic gain is approximately $\sqrt{2mr}$. We give a few examples of the time and memory complexities of various $Dissect_u(r,m)$ algorithms in Table 1 in the Appendix. An analysis presented in Appendix A.3 shows that the gain obtained by this sequence is optimal amongst dissection algorithms with 2^{mn} memory.

A.3 Analysis of the Algorithm $Dissect_u(r, m)$

Similarly to $Dissect_u(r, 1)$, the $Dissect_u(r, m)$ algorithm on r-encryption is defined as follows:

- 1. Given r plaintext-ciphertext pairs $(P_1, C_1), (P_2, C_2), \dots (P_r, C_r)$, for each possible value of $X_1^u, X_2^u, \dots, X_{u-m}^u$ perform:
 - (a) Obtain the 2^{mn} candidates keys for k_1, k_2, \ldots, k_u given the plaintext-"ciphertext" pairs $(P_1, X_1^u), (P_2, X_2^u), \ldots, (P_{u-m}, X_{u-m}^u)$. For each such candidate key compute $E^{[1...u]}(P_{u-m+1}, P_{u-m+2}, \ldots)$ and store the outcome in a table (along with the corresponding key candidate k_1, k_2, \ldots, k_u).

(b) Obtain the $2^{[r-u-(u-m)]n}$ candidate keys⁹ $k_{u+1}, k_{u+2}, \ldots, k_r$, and compute $D^{[u+1...r]}(C_{u-m+1}, C_{u-m+2}, \ldots)$ for each such candidate and check whether the corresponding value is in the table. If so, check the key candidate offered by the value of the current candidate of $k_{u+1}, k_{u+2}, \ldots, k_r$ with the value stored in the table for k_1, k_2, \ldots, k_u .

Analysis of $Dissect_u(r, m)$: As for $Dissect_u(r, 1)$, it is easy to see that each of the steps of the algorithm uses at most 2^{mn} memory if the sub-attacks of Step 2(a) and Step 2(b) each uses 2^{mn} memory. When running the attacks of Steps 2(a) and Steps 2(b), obviously we need to call the optimal attacks themselves, yielding the optimal attacks for $Dissect_u(r, m)$.

As before, we note that all complexities involve the n factor in the exponent (e.g., 2^{mn} memory) we define by $f_m(r)$ the exponent divided by n of the time complexity of the optimal attack¹⁰ on r-encryption with 2^{mn} memory. In other words, the optimal attack on r-encryption with 2^{mn} memory takes $2^{f_m(r)n}$ time.

The time complexity of $Dissect_u(r, m)$ is equal to that of $2^{(u-m)n}$ repetitions of $Dissect_{u'}(u, m)$ and $Dissect_{u^*}(r-u, m)$ plus $2^{(r-2u+m)n}$ accesses to a table. Obviously, we suggest picking u' and u^* optimally, and thus, the time complexity of $Dissect_u(r, m)$ is

$$2^{(u-m)n} \cdot \max \left\{ 2^{f_m(u)n}, 2^{f_m(r-u)n}, 2^{(r-2u+m)n} \right\}.$$

Therefore, for a given value of r the optimal value of u is the one that minimizes the above expression. In other words:

$$f_m(r) = \min_{1 \le u \le r} \left\{ u - m + \max \left\{ f_m(u), f_m(r - u), r - 2u + m \right\} \right\}.$$
 (4)

Using Equation (4), it is easy to show by induction on r that the minimal complexity is achieved by the sequence of algorithms presented in Section A.2.

We conclude by offering in Table 1 values of $f_m(r)$ for several choices of m and r.

A.4 Dissection Algorithms for a Composition of One-Way Functions

In this section we consider compositions of one-way functions (OWFs), which appear in the context of layered Message Authentication Codes, such as NMAC [2]. It turns out that the deterministic dissection algorithms described in Section 3 can be easily modified to yield optimal dissection algorithms for this scenario. We note that the algorithms based on parallel collision search, presented in Section 4 cannot be applied in this scenario (for $m = 2^n$), since they make crucial use of decryption.

⁹ We note that when attacking $E^{[u+1...n]}$, we expect that out of the $2^{(r-t)n}$ possible keys, only one in $2^{-(u-m)n}$ is consistent with the given given u-m "plaintext"-ciphertext pairs.

¹⁰ Optimal in that sense means lowest time complexity.

Enc.	m = 1	m=2	m=3	m=4	m=5	m = 6	m = 7	m = 8	m = 9	m = 10
1	1	1	1	1	1	1	1	1	1	1
2	1	1	1	1	1	1	1	1	1	1
3	2	2	2	2	2	2	2	2	2	2
4	2	2	2	2	2	2	2	2	2	2
5	3	3	3	3	3	3	3	3	3	3
6	4	3	3	3	3	3	3	3	3	3
7	4	4	4	4	4	4	4	4	4	4
8	5	4	4	4	4	4	4	4	4	4
9	6	5	5	5	5	5	5	5	5	5
10	7	6	5	5	5	5	5	5	5	5
11	7	6	6	6	6	6	6	6	6	6
12	8	7	6	6	6	6	6	6	6	6
13	9	8	7	7	7	7	7	7	7	7
14	10	8	8	7	7	7	7	7	7	7
15	11	9	8	8	8	8	8	8	8	8
16	11	10	9	8	8	8	8	8	8	8
17	12	11	10	9	9	9	9	9	9	9
18	13	11	10	10	9	9	9	9	9	9
19	14	12	11	10	10	10	10	10	10	10
20	15	13	12	11	10	10	10	10	10	10
21	16	14	12	12	11	11	11	11	11	11
22	16	14	13	12	12	11	11	11	11	11
23	17	15	14	13	12	12	12	12	12	12
24	18	16	15	14	13	12	12	12	12	12
25	19	17	15	14	14	13	13	13	13	13
26	20	18	16	15	14	14	13	13	13	13
27	21	18	17	16	15	14	14	14	14	14
28	22	19	18	16	16	15	14	14	14	14
29	22	20	18	17	16	16	15	15	15	15
30	23	21	19	18	17	16	16	15	15	15
31	24	22	20	19	18	17	16	16	16	16
32	25	22	21	19	18	18	17	16	16	16
33	26	23	21	20	19	18	18	17	17	17
34	27	24	22	21	20	19	18	18	17	17
35	28	25	23	22	20	20	19	18	18	18
36	29	26	24	22	21	20	20	19	18	18
37	29	27	25	23	22	21	20	20	19	19
38	30	27	25	24	23	22	21	20	20	19
39	31	28	26	25 25	23	22	22	21	20	20
40	32 Table	29	27	25	24	23	22	22	21	20

Table 1. $f_m(r)$ Values. Items marked in bold are magic numbers.

In the scenario of composition of OWFs, the problem at end is to retrieve k_1, k_2, \ldots, k_r used in

$$F^{[1...r]}(P) = F_{k_r}(F_{k_{r-1}}(\cdots F_{k_2}(F_{k_1}(P))\cdots)),$$

where $F_k(\cdot)$ is a one way function from *n*-bit input and *n*-bit key into *n*-bit output.

Given 2^{2n} memory, one can simply store a table of $(F_k(X), k, X)$ sorted by the values of $F_k(X)$ and k. Thus, given $F_k(X)$ and k, one can find all possible X values (there is one such value on average) that are indeed "encrypted" into $F_k(X)$ under the key k. As this actually generates the "decryption" functionality by one memory access per each possible X (and thus amortized complexity of one memory access per each $F_k(X)$, k tuple), we can repeat the same $Dissect_u(r, m)$ algorithms as for encryption if $m \geq 2$.

Once we are given only a restricted amount of 2^n memory, we use a slightly different approach to the dissection of $F^{[1...r]}$ into two parts. Instead of guessing intermediate values and checking which keys "decrypt" the known outputs¹¹ into these intermediate values, we guess the intermediate values and try the keys, until we find keys for which the intermediate value is correct. This procedure is somewhat less efficient than the original $Dissect_u(r, m)$, but obtains similar results.

Consider the $DissectOWF_2(3)$ algorithm (recall that we discuss only the case for which 2^n memory is given, as for 2^{2n} memory the previous Dissect algorithms are still applicable):

- 1. For each possible value of X_1^2 :
 - (a) For each possible k_3 , check whether $F_{k_3}(X_1^2) = C_1$. For each k_3 which satisfies this relation store in a table the value X_1^2, k_3 .
- 2. For every k_1, k_2 compute $F_{k_2}(F_{k_1}(P_1))$ and check whether the obtained value is in the table. If so, test the key suggested by the collision by checking whether $F^{[1...3]}(P_2, P_3) = C_2, C_3$.

Step 1 goes over 2^{2n} values, with an n-bit filtering condition. Hence, its running time is 2^{2n} and the expected memory consumption is 2^{2n} . The time complexity of Step 2 is 2^{2n} , and for each key pair k_1, k_2 we expect on average one value of X_1^2 in the table (that suggests on average a single k_3 for a complete key suggestion. We conclude that the time complexity of the $DissectOWF_2(3)$ is 2^{2n} and its memory complexity is 2^n .

We note that a trivial extension of $DissectOWF_2(3)$ allows to retrieve the key of a composition of r OWFs for any $r \geq 3$, in time $2^{(r-1)n}$ and memory 2^n . Moreover, if we are given only r-1 input/output pairs, the same algorithm

¹¹ As we attack one way functions we use the terms inputs and outputs rather than plaintexts and ciphertexts.

A given output may have several inputs (even when the key is fixed), thus the number of "solutions" to the equation $F_{k_3}(X_1^2) = C_1$ may be higher than 2^n . However, assuming $F(\cdot)$ is a "good" one way function, the number of solutions is not expected to be significantly higher than 2^n .

allows to retrieve the 2^n keys which comply with these pairs. This algorithm will be used in the recursive step below.

Starting with $DissectOWF_2(3)$, we can construct recursively a sequence of dissection algorithms. The construction is similar to the $Dissect_{NEXT}$ construction presented in Section 3, with a few difference which follow from the special structure of OWFs.

Given $\ell \geq 2$ and an algorithm $DissectOWF_j(r')$ such that $Gain(DissectOWF_j(r')) = \ell - 1$, we define the algorithm $DissectOWF_{NEXT} = DissectOWF_{r'}(r' + \ell + 1)$ for r-encryption, where $r = r' + \ell + 1$, as follows:

- 1. Given r plaintext-ciphertext pairs $(P_1, C_1), (P_2, C_2), \dots (P_r, C_r)$, for each possible value of $X_1^{r'}, \dots, X_\ell^{r'}$, perform:
 - (a) Apply the basic extension of the $DissectOWF_2(3)$ algorithm to $E^{[r'+1...r]}$ with $(X_1^{r'}, C_1), \ldots, (X_\ell^{r'}, C_\ell)$ as the plaintext/ciphertext pairs, and obtain 2^n candidates for the keys $(k_{r'+1}, ldots, k_r)$. For each such candidate, compute the preimages $(X_{\ell+1}^{r'}, \ldots, X_r^{r'})$ of $(C_{\ell+1}, \ldots, C_r)$, by going over all possible inputs to $E^{[r'+1...r]}$, encrypting them with the $(k_{r'+1}, \ldots, k_r)$ candidate, and storing the inputs whose corresponding outputs are $(C_{\ell+1}, \ldots, C_r)$. Store them in a table, along with the corresponding key candidate $(k_1, \ldots, k_\ell + 1)$.
 - (b) Apply the algorithm $DissectOWF_j(r')$ to $E^{[1...r']}$ with $(P_1, X_1^{r'}), \ldots, (P_\ell, X_\ell^{r'})$ as the plaintext/ciphertext pairs. Check each of the $2^{(r'-\ell)n}$ suggestions for the keys $(k_1, \ldots, k_{r'})$ on-the-fly by partially encrypting $P_{\ell+1}, \ldots, P_r$, and checking whether the corresponding vector $(X_{\ell+1}^{r'}, \ldots, X_r^{r'})$ appears in the table.

An analysis similar to that of the $Dissect_{NEXT}^1$ algorithm presented in Section 3.4 shows that the time and memory complexities of $DissectOWF_{NEXT}$ are $T=2^{r'n}$ and $M=2^n$. Indeed, the only essential difference between $DissectOWF_{NEXT}$ and $Dissect_{NEXT}^1$ is the second part of Step 2(a) (i.e., computing the preimages), but the time complexity of this part is 2^{2n} , which is less than the complexity of other steps of the algorithm since $\ell \geq 2$.

This shows that the sequence of $DissectOWF_j(r)$ algorithms satisfies the same recursion relation as the sequence $Dissect_j(r, 1)$. Hence, if we denote the number r of rounds in the ℓ 's element of the sequence (i.e., the element for which the gain equals to ℓ) by \tilde{r}_{ℓ} , then

$$\tilde{r}_{\ell} = \tilde{r}_{\ell-1} + \ell + 1.$$

Since $\tilde{r}_1 = 5$ (which can be easily checked manually), we get the formula

$$\tilde{r}_{\ell} = r_{\ell} + 1 = \frac{\ell(\ell+1)}{2} + 2.$$

Therefore, the "magic sequence" corresponding to a composition of OWFs is

$$Magic_1^{OWF} = \{2, 3, 5, 8, 12, 17, 23, \ldots\},\$$

and the asymptotic gain is approximately $\sqrt{2r}$, as for the basic r-encryption case.

A.5 The Improved Rebound Attack

Our dissection algorithms can be applied for this problem as well, replacing gradual matching or parallel matching suggested in [13]. Consider for example the rebound attack on Luffa. In this rebound attack, the adversary is given 2^{67} possible input differences ($|L_A| = 2^{67}$) and $2^{65.6}$ output differences ($L_B = 2^{65.6}$), and has to find valid input/output differences from the S-box layer. All the differences are 208-bit strings, as there are 52 active 4-bit to 4-bit S-boxes. In each such S-box, for any given input difference there are about 7 possible output differences (and vice versa). For space reasons we refer the reader to [13] for the description of the currently best algorithm which takes time of 2^{104} and 2^{102} memory.

Using our dissection algorithm we offer an improved attack with a significantly reduced memory complexity:

- 1. For each input difference X to the first 13 S-boxes:
 - (a) For each difference $\delta \in L_A$ which agrees with X on the first 13 S-boxes, store in a table all possible differences that can be caused by the reminder of δ in the next 13 S-boxes (along with the value of δ).
 - (b) For each possible difference that X may cause through the S-box layer (in the first 13 S-box), for each difference $\delta' \in L_B$ that agrees with this difference over the first 13 S-boxes, check whether their difference in the next 13 S-boxes appears in the table. If so, check whether δ and δ' can match over the remaining S-boxes.

An analysis of this algorithm shows that its time complexity is 2^{104} operations, but its memory complexity is only 2^{52} . We note that the true memory complexity of this attack is actually 2^{66} , as one needs to store at least one out of L_A or L_B . Still, our results significantly improves the ones of [13].

We note that this algorithm, besides improving other rebound attacks, can also be used when the problem is composed of layers which are relations (rather than functions or permutations) which allow multiple outputs for a single input.

B The Knapsack Reduction and its Analysis

The knapsack problem is defined as follows: given a list of n positive integers a_1, a_2, \ldots, a_n of n bits and an additional n-bit positive integer S, find a vector $\epsilon = (\epsilon_1, \epsilon_2, \ldots, \epsilon_n)$, where $\epsilon_i \in \{0, 1\}$, such that $S = \sum_{i=1}^n \epsilon_i \cdot a_i \pmod{2^n}$.

B.1 Reduction of the Knapsack Problem to Multiple Encryption

We show how to reduce the knapsack problem to a multiple encryption problem with r rounds, for any positive integral $r \ll n$: Our reduction treats the unknown

¹³ We work with modular knapsacks which are in general computationally equivalent to arbitrary knapsacks [6].

 ϵ as an *n*-bit key of a block cipher: given an *n*-bit "plaintext" P, the "ciphertext" is defined as $P + \sum_{i=1}^{n} \epsilon_i \cdot a_i \pmod{2^n}$. Similarly, given an n-bit ciphertext C, we define the inverse decryption function as the computation of $C - \left(\sum_{i=1}^{n} \epsilon_i \cdot a_i\right) \pmod{2^n}$. Thus, by fixing the plaintext to the *n*-bit vector $\mathbf{0}$, and defining the ciphertext as S, the knapsack problem reduces to recovering the key of this block

We now show how to reduce the knapsack problem to an r-encryption algorithm:

cipher, given one plaintext-ciphertext pair.

- We write the knapsack as a sum of r knapsacks: $\sum_{i=1}^r \sigma^i = S \pmod{2^n}$, where the i'th knapsack for $i \in \{1,2,\ldots,r\}$ is defined as $\sigma^i = \sum_{j=1+in/r}^{(i+1)n/r} \epsilon_j \cdot a_j \pmod{2^n}$. (mod 2^n). Correspondingly, we split the key ϵ into n/r-bit "round keys", where the i'th round key for $i \in \{1,2,\ldots,r\}$ is defined as $\epsilon_{[1+in/r,2+in/r,\ldots,(i+1)n/r]}$ (bits $[1+in/r,2+in/r,\ldots,(i+1)n/r]$ of ϵ). Splitting the knapsack can thus be viewed as splitting the "block cipher" **horizontally** into smaller encryption rounds with n/r-bit keys, where $X^i \triangleq \sum_{j=1}^i \sigma^j$ for $i \in \{1,2,\ldots,r-1\}$ is an n-bit intermediate encryption value.
- Our algorithms work on block ciphers where the block size is equal to the key size, which allows them to guess (and to partially encrypt and decrypt) intermediate encryption values at a much lower cost that guessing the full key. Thus, we must also split the "block cipher" **vertically** into n/r-bit blocks: we split the n-bit zero plaintext into r smaller zero n/r-bit plaintexts. The corresponding ciphertexts are obtained by splitting the n-bit ciphertext S into r smaller n/r-bit ciphertexts, where the j'th ciphertext is defined as $S_{[1+jn/r,2+jn/r,...,(j+1)n/r]}$. Similarly, for the j'th "plaintext-ciphertext" pair, the i'th intermediate encryption value is defined by taking the j'th block of bits of X^i , namely $X^i_{[1+jn/r,2+jn/r,...,(j+1)n/r]}$.
- bits of X^i , namely $X^i_{[1+jn/r,2+jn/r,...,(j+1)n/r]}$.

 In total, we decompose of the knapsack problem both horizontally and vertically into r^2 small knapsacks, where each small knapsack has an horizontal index $i \in \{1, 2, ..., r\}$ and a vertical index $j \in \{1, 2, ..., r\}$. The reduction views knapsack(i, j) as an n/r-bit block cipher which encrypts an n/r-bit "plaintext" P using an additional carry value c_{i_j} . The **encryption func-**

tion (i,j) is defined as $P + c_{i_j} + \sum_{i=1+in/r}^{(i+1)n/r} \epsilon_i \cdot a_i \pmod{2^{n/r}}$. The next carry value $c_{i_{j+1}}$ is defined as the carry value of this addition operation modulo $2^{n/r}$, and c_{i_0} is equal to zero for all $i \in \{1, 2, \ldots, r\}$.

 $[\]overline{^{14}}$ Since we are mostly interested in asymptotic analysis, we assume for the sake of simplify that n is divisible by r.

¹⁵ We also define the inverse decryption function.

A potential problem with this formulation is that unlike the standard case of multiple encryption, here the encryption of each plaintext P_i depends on the encryption of P_{i-1} via the carry values $c_{i_1}, c_{i_2}, \ldots, c_{i_{r-1}}$. In order to avoid this problem, the underlying multiple encryption algorithm simply guesses the intermediate encryption values in their natural order (i.e., whenever we need to guess several sequences of n/r bits of some X^i , we guess them as a continuous sequence of bits, starting at the LSB). Thus, we can keep track of the various carry values for free without having to guess them separately.

B.2 Optimal Parameters for the Knapsack Reduction

Recall from Section 5.1 that given r, we apply the multiple encryption algorithm with an effective block size of $n^* = n/r$ and an effective memory unit of $m^* = mr$. We would like to find r that minimizes the running time of our algorithm $f(r, n^*, m^*) = f(r, n/r, mr)$.

For any integral value z, it is easy to reduce our multiple encryption algorithm with parameters $(z \cdot r, n^*/z, z \cdot m^*)$, to an algorithm with parameters (r, n^*, m^*) that runs with the same time complexity. As described in Section 3.4, the reduction aggregates every sequence of z blocks into a single block, and every sequence of z plaintext-ciphertext pairs into a single pair. The reduction implies that for any integral value of z, $f(z \cdot r, n^*/z, z \cdot m^*) \leq f(r, n^*, m^*)$. In view of the knapsack problem, this implies that given n and m, in order to find an optimal value of r, it is sufficient to consider values of r which are multiples of some integer. Thus, for the sake of simplicity, we choose to consider only values of r for which $m^* = rm$ is integral. We note that an optimal value of r is clearly not unique, since if r is optimal it implies that $r^* = zr$ is also optimal for any positive integral z.

Deterministic Algorithms We analyze the algorithm which solves the knap-sack problem using our deterministic multiple encryption algorithms constructed in Section 3.

A Few Examples Before considering the general problem, we find optimal values of r in a few examples: for m=1/4, we consider $r\in\{4,8,12,\ldots\}$ starting with r=4 and $m^*=4\cdot(1/4)=1$. In this case, we need to consider the series $\{2,4,7,11,\ldots\}$ and apply our 4-round algorithm, for which the improvement factor over exhaustive search is $2^{2m^*n^*}=2^{2n/4}=2^{n/2}$. For r=8, we get $m^*=2$. In this case, we need to consider the series $\{4,6,8,11,14,18,22,\ldots\}$ and apply our 8-round algorithm, for which the improvement factor over exhaustive search is $2^{4m^*n^*}=2^{4n/8}=2^{n/2}$. Similarly, it is easy to check that for every $r\in\{4,8,12,\ldots\}$, we get the same improvement factor of $2^{n/2}$. Thus, in the case of m=1/4, choosing $r\in\{4,8,12,\ldots\}$ does not affect the time complexity of the knapsack algorithm, which also implies that any choice of $r\in\{4,8,12,\ldots\}$ is optimal.

As shown above, for m = 1/4 the complexity of the attack does not depend on values of r (for which $m^* = rm$ is integral). However, this is not always the case. For example, for m = 1/9 we consider $r \in \{9, 18, 27, \ldots\}$: for r = 9, we get $m^* = 1$ and consider the series $\{2, 4, 7, 11, \ldots\}$. Thus, our algorithm is determined by the 7-round algorithm (after guessing 2 round keys), for which the improvement factor over exhaustive search is $2^{3m^*n^*} = 2^{3n/9}$. For r = 18, we get $m^* = 2$ and consider the series $\{4, 6, 8, 11, 14, 18, 22, \ldots\}$. Thus, our algorithm is determined by the 18-round algorithm (without guessing any round keys), for which the improvement factor over exhaustive search is $2^{7m^*n^*}=2^{7n/18}$. For r = 27, we get $m^* = 3$, consider the series $\{6, 8, 10, 12, 15, 18, 21, 25, 29, 33, \ldots\}$ and get an improvement factor of $2^{11n/27}$. Similarly to r=18, for r=36, we get an improvement factor of $2^{14n/36} = 2^{7n/18}$. Since $2^{7n/18} > 2^{11n/27} > 2^{3n/9}$, then for m = 1/9 it is more efficient to divide the knapsack with r = 18 (or r = 36) knapsacks than with r = 9 or r = 27 knapsacks. For the case of m = 1/9, it is possible to show that for any positive r which is a multiple of 18, our knapsack algorithm has the same time complexity. Thus, choosing r = 18z for a positive integral z optimizes the algorithm.

Optimal Parameters for the General Case of Deterministic Algorithms

More generally, given m, a sufficient condition for the optimality of r is that it belongs to the series constructed in Section 3.4, namely $r \in \{2m^* + 2i, 4m^* + 3i, 7m^* + 4i, 11m^* + 5i, 16m^* + 6i, \ldots\}$ for some integral $0 \le i < m^* = mr$). Intuitively, this is sufficient because when $r \in \{2m^* + 2i, 4m^* + 3i, \ldots\}$, the algorithm does not require any key guesses, and achieves the optimal improvement factor. Note that this condition holds for the optimal values of r that we calculated for $m \in \{1/4, 1/8, 1/9\}$.

Given a fixed value of m we would like to compute a value of r for which the sufficient condition for optimality holds. However, we first show that the condition on r is indeed sufficient for optimality by showing that the knapsack algorithm has the same running time for any integral multiple of $r \in \{2m^* + 2i, 4m^* + 3i, \ldots\}$: write $r = b_{j-1}m^* + ji$ for $0 \le i < r$ (where b_j denoted the j'th element of the "magic sequence" constructed in Section 3.4, starting from j = 0), and take r' = zr for some positive integral z which gives parameters $m' = zm^*$ and $n' = m^*/z$ for the r'-encryption algorithm. We have $r' = zr = b_{j-1}zm^* + jzi = b_{j-1}m' + j(zi)$, for $0 \le iz < r'$, and thus the improvement factor of the r'-round algorithm over exhaustive search is $2^{(j-1)m'+zi)n'} = 2^{((j-1)zm^*+zi)m^*/z} = 2^{((j-1)m+i/r)n}$, which is indeed independent of z.

According to our sufficient condition for optimality of r, given m we would like to optimize the knapsack algorithm by finding a value of r such that $m^* = mr$ is integral and $r = b_{j-1}mr + ji$, for some integral j and $0 \le i < mr$. In order to find such a value of r, we find the value of j for which $b_{j-1} \le 1/m < b_j$ (or $b_{j-1} \le 1/m < b_{j-1} + j$), and assume that m = p/q for some integral p and q. We now show that r = qj satisfies the condition for optimality: first, $m^* = mr = pj$ is indeed integral. Second, $r - b_{j-1}mr = qj - b_{j-1}pj = j(q - b_{j-1}p)$, and we need to show that $0 \le i \triangleq q - b_{j-1}p < mr$. Indeed, since $b_{j-1} \le q/p < b_{j-1} + j$,

then $b_{j-1}p \le q < b_{j-1}p + jp$, implying $0 \le q - b_{j-1}p < jp = mr$. Thus, r = qj is a value of r for which our knapsack algorithm is optimal.

We note that for a given value of m=p/q such that $b_{j-1} \leq 1/m < b_j, r=qj$ is not necessarily the smallest number of rounds for which we get an optimal algorithm (even when the greatest common divisor of p and q is 1). For example, in case m=1/4, we have $4 \leq 4 < 7$ ($b_{j-1}=4$ and $b_{j-1}+j=7$), i.e, j=3 and r=12. However, we can also get the optimal value by taking r=4. In case m=1/9, we have $7 \leq 9 < 11$, i.e, j=4 and r=36. However, we can also get the optimal value by taking r=1/8, we have $7 \leq 8 < 11$, i.e, j=4, r=32. It can be easily verified that for m=1/8, r=32 is the smallest value of r which gives the optimal knapsack algorithm.

The Time Complexity of the Resultant Knapsack Algorithm Next, we analyze the time complexity of the knapsack algorithm for r=qj: since we have $r=b_{j-1}m^*+ji$ for $i=q-b_{j-1}p$, the improvement factor of the algorithm over exhaustive search is $2^{((j-1)m^*+i)n^*}=2^{((j-1)mr+i)n/r}=2^{((j-1)m+i/r)n}=2^{((j-1)m+(q-b_{j-1}p)/qj)n}=2^{((j-1)m+(1-b_{j-1}m)/j)n}=2^{((j-1-b_{j-1}/j)m+1/j)n}$. Recall from Appendix A.2 that $b_j=\frac{j^2+j+2}{2}$, or $b_{j-1}/j=(j-1+2/j)/2$. Plugging in this expression into our improvement factor, we obtain $2^{(\frac{j-1-2/j}{2}m+\frac{1}{j})n}$. For example, for $4\leq 1/m<7$, we have j=3 and obtain an improvement factor of $2^{(\frac{2-2/3}{2}m+\frac{1}{3})n}=2^{(\frac{2}{3}m+\frac{1}{3})n}$ (e.g., for m=1/4, we get an improvement factor of $2^{0.5n}$).

We can now write the running time of the knapsack algorithm as $T = 2^{tn}$, where $t = 1 - (\frac{j-1-2/j}{2}m + \frac{1}{j})$. When $b_{j-1} \leq 1/m < b_j$, j is fixed to a constant value and consequently t is a linear function of m. Thus, our deterministic algorithms connect in a straight line any two consecutive time-memory tradeoff points (i.e, points with a constant value of j) for $1/m \in \{4, 7, 11, 16, \ldots\}$, and we obtain a continuous curve (as claimed in Section 5.1).

C Further Analysis of the Dissect & Collide Algorithms

C.1 The Parallel Collision Search (PCS) Algorithm

In this section we describe the Parallel Collision Search (PCS) algorithm of van Oorschot and Wiener [15]. For more information and the algorithm and its applications, the reader is referred to [15].

The Memoryless Algorithm The simplest way to present the PCS algorithm is to consider "memoryless" (i.e., constant memory) attacks on r-encryption. As mentioned in Section 3, the time complexity of exhaustive search is 2^{rn} , and the meet in the middle attack does not perform better if only a constant memory is allowed. Van Oorschot and Wiener showed that the time complexity can be reduced to $2^{(3/4)rn}$, using the PCS algorithm.

The basic observation behind the algorithm is that one can find, efficiently and with only a constant memory, key candidates which comply with half of the plaintext/ciphertext pairs.

Assume, for sake of simplicity, than r is even, and the adversary is given r plaintext/ciphertext pairs $(P_1, C_1), \ldots, (P_r, C_r)$. The first step of the PCS algorithm consists of finding candidates of the keys (k_1, \ldots, k_r) , which comply with the pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$. In order to find them, the adversary constructs two step functions:

$$F^{upper}: (k_1, \dots, k_{r/2}) \mapsto (X_1^{r/2}, \dots, X_{r/2}^{r/2})$$
 and $F^{lower}: (k_{r/2+1}, \dots, k_r) \mapsto (X_1^{r/2}, \dots, X_{r/2}^{r/2}),$

and uses Floyd's cycle finding algorithm [8] to find a collision between them. It is clear that both functions can be computed efficiently given the pairs $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$, and that a collision between them yields a value of $(k_1, \ldots, k_{r/2}, k_{r/2+1}, \ldots, k_r)$ which complies with $(P_1, C_1), \ldots, (P_{r/2}, C_{r/2})$. As both functions are from (r/2)n bits to (r/2)n bits, Floyd's algorithm is expected to find a collision in time $2^{(r/4)n}$, with constant memory. In the sequel, we call such collisions partial collisions.

In the second step of the algorithm, the adversary checks whether the found key candidate complies also with the pairs $(P_{r/2+1}, C_{r/2+1}), \ldots, (P_r, C_r)$. By a standard randomness assumption, this occurs with probability $2^{-(r/2)n}$, and hence, it is expected that after $2^{(r/2)n}$ candidates, a right key candidate is found. The total time complexity of the algorithm is thus $2^{(3/4)rn}$.

It should be noted that the adversary has to use different flavors of the step functions F^{upper} and F^{lower} in order to produce the $2^{(r/2)n}$ distinct partial collisions required for the second step of the algorithm. In addition, the algorithm is probabilistic, in the sense that its success probability depends on the (randomly chosen) starting points of Floyd's algorithm.

A Time/Memory Tradeoff If more memory is available, then the algorithm described above can be combined with the techniques used in the classical Hellman's time-memory tradeoff attack [4] to obtain the tradeoff curve $T^2M = 2^{(3/2)rn}$. The reduction in the time complexity is in the first step of the attack (i.e., obtaining the partial collisions), and is achieved by obtaining many partial collisions simultaneously.

Assume that the available memory is $M=2^{mn}$. The adversary chooses M random starting points V_i , and for each of them she computes the sequence $\{V_i, F^{upper}(V_i), F^{upper}(F^{upper}(V_i)), \ldots\}$. Each sequence is terminated once a value with (r/4-m/2)n opening zeros is obtained, and this "distinguished point" is stored in a table. Then the adversary performs the same operation with the step function F^{lower} , and for any reached distinguished point, she checks whether it appears in the table. If it appears, then the respective paths of F^{upper} and F^{lower} collide, and the collision can be found by Floyd's algorithm.

As the expected length of each path is $2^{(r/4-m/2)n}$ and there are 2^{mn} paths, it is expected that each path corresponding to F^{upper} collides with one path

corresponding to F^{lower} on average. Hence, in total, 2^{mn} partial collisions exist in the examined data.

The generation of the table and the accesses to it require $2 \cdot 2^{mn} \cdot 2^{(r/4-m/2)n} \approx 2^{(r/4+m/2)n}$ operations. Each of the 2^{mn} partial collisions is then found by running Floyd's algorithm on two colliding paths, which requires time of $2^{(r/4-m/2)n}$ (since this is the expected length of the paths). Hence, this step also requires $2^{(r/4+m/2)n}$ operations.

Therefore, the algorithm requires about $2^{(r/4+m/2)n}$ operations and finds 2^{mn} partial collisions. Since $2^{rn/2}$ partial collisions are needed, the overall time complexity is

$$T = 2^{(r/4+m/2)n} \cdot 2^{(r/2-m)n} = 2^{(3r/4-m/2)n}$$

which leads to the tradeoff curve $T^2M = 2^{(3/2)rn}$.

C.2 Flavors in the Step Function of the Algorithm DC

In this appendix we discuss an important difference between the DC algorithm and the PCS algorithm. In PCS, for each value of $(k_1,\ldots,k_{r/2})$, there is exactly one corresponding value of $F^{upper}(k_1,\ldots,k_{r/2})=(X_1^{r/2},\ldots,X_{r/2}^{r/2})$. In DC, for some values of I^{upper} , there are no corresponding values of $\tilde{F}^{upper}(I^{upper})=(X_{u+1}^{r/2},\ldots,X_{r/2}^{r/2})$ at all, while for other values of I^{upper} , there are several possible outputs of \tilde{F}^{upper} . This happens since the number of keys $(k_1,\ldots,k_{r/2})$ which comply with the u plaintext/ciphertext values $(P_1,X_1^{r/2}),\ldots,(P_u,X_u^{r/2})$ and the r/2-u fixed intermediate values contained in I^{upper} , is not constantly 1, but is distributed according to a Poisson distribution with mean 1. While this feature does not influence the average performance of the Dissect attacks, its effect on the DC attack is crucial: in an e^{-1} fraction of the cases, the step function \tilde{F}^{upper} returns no value, and thus, the expected length of the paths generated by it is constant!

In order to resolve this difficulty, we introduce *flavors* into the definition of the step function. Formally, for each value of I^{upper} , $\tilde{F}^{upper}(I^{upper})$ is a (possibly empty) multiset. Based on this, we define:

$$\bar{F}^{upper}(I^{upper}) = \min(\{\tilde{F}^{upper}(I^{upper} \oplus i_0)\}),$$

where $i_0 \in \{0,1\}^{(r/2-u)n}$ is minimal such that the set $\{\tilde{F}^{upper}(I^{upper} \oplus i_0)\}$ is non-empty, and the minimums are taken with respect to the lexicographic order. In other words, if the set $\tilde{F}^{upper}(I^{upper})$ is empty, then we replace I^{upper} by $I^{upper} \oplus (0,0,\ldots,0,1)$ and compute \tilde{F}^{upper} again. We continue until we reach a value of i_0 for each the set of outputs of \tilde{F}^{upper} is non-empty, and then we choose the least output (in the lexicographic order) to be the output $\bar{F}^{upper}(I^{upper})$. The same modification is applied also to \tilde{F}^{lower} .

Using this modification, the step function becomes uniquely defined like in the case of PCS, and the computation overhead required by the new definition is small, since the output of \tilde{F}^{upper} is non-empty for an 1-1/e fraction of the

inputs. It can be shown that the success probability of the modified variant of the DC algorithm is similar to the success probability of the PCS algorithm. Finally, we note that the PCS algorithm also uses different flavors of the functions F^{upper} and F^{lower} in order to get a sufficiently large number of distinct partial collisions (see Appendix C.1). However, while in PCS the flavors are changed only after some partial collisions are found, in DC the flavors must be part of the step function.

C.3 The Monotonicity of the Gain of DCAlgorithms

In this subsection we prove that the gain of the DC algorithms over the PCS algorithm is monotone non-decreasing, as a function of r.

Lemma 2. Assume that an algorithm A of the class DC(r', m) has gain ℓ . Then there exists an algorithm B of the class DC(r'+2, m) whose gain is also equal to ℓ .

Proof. Recall that the sets of intermediate values fixed in the algorithm A are denoted by $I^{upper}(A)$ and $I^{lower}(A)$. We specify the algorithm B by fixing the sets of intermediate values: $I^{upper}(B) = I^{upper}(A) \cup \{X_1^{r'/2}\}$, and similarly for $I^{lower}(B)$. Note that as the DC algorithms for r-encryption satisfy $u + |I^{upper}| = r$, our choice of $I^{upper}(B)$ and $I^{lower}(B)$ ensures that u(B) = u(A). Hence, Equation (1) implies that in order to show that $Gain_{ND}(B) = Gain_{ND}(A)$, it is sufficient to show that s(B) = s(A). Let Step(A) be an algorithm which allows to compute the function $\tilde{F}^{upper}(A) : I^{upper}(A) \mapsto (X_{u+1}^{r'/2}, \dots, X_{r'/2}^{r'/2})$ in time 2^{sn} (where u = u(A) and s = s(A)), given the plaintext/ciphertext pairs $(P_1, X_1^{r'/2}), \dots, (P_u, X_u^{r'/2})$ and the intermediate values contained in $I^{upper}(A)$. We have to find an algorithm Step(B) which computes the function $\tilde{F}^{upper}(B) : I^{upper}(B) \mapsto (X_{u+1}^{(r'+2)/2}, \dots, X_{(r'+2)/2}^{(r'+2)/2})$ in time 2^{sn} given the plaintext/ciphertext pairs $(P_1, X_1^{(r'+2)/2}), \dots, (P_u, X_u^{(r'+2)/2})$ and the intermediate values contained in $I^{upper}(B)$. We define the algorithm Step(B) as follows:

- 1. Use the values $X_1^{r'/2}$ and $X_1^{(r'+2)/2}$ to compute a unique value of the key $k_{(r'+2)/2}$ which complies with them.
- 2. Use the value of $k_{(r'+2)/2}$ to partially decrypt the vector $(X_1^{(r'+2)/2}, \ldots, X_u^{(r'+2)/2})$ through $E^{(r'+2)/2}$ to obtain the vector $(X_1^{r'/2}, \ldots, X_u^{r'/2})$.
- through $E^{(r'+2)/2}$ to obtain the vector $(X_1^{r'/2}, \ldots, X_u^{r'/2})$.

 3. Use the algorithm Step(A) to deduce the vector $(X_{u+1}^{r'/2}, \ldots, X_{r'/2}^{r'/2})$ from the plaintext/ciphertext pairs $(P_1, X_1^{r'/2}), \ldots, (P_u, X_u^{r'/2})$ and the intermediate values contained in $I^{upper}(A)$.
- 4. Partially encrypt the vector $(P_1, X_1^{r'/2}), \ldots, (P_u, X_u^{r'/2})$ through $E^{(r'+2)/2}$ to obtain the desired vector $\tilde{F}^{upper}(B)(I^{upper}(B)) = (X_{u+1}^{(r'+2)/2}, \ldots, X_{(r'+2)/2}^{(r'+2)/2})$.

It is clear that Step 1 of the algorithm requires at most 2^n operations, Steps 2 and 4 require at most r operations each, and Step 3 requires 2^{sn} operations.

Hence, if $s \geq 1$ (which is the case for all DC algorithms), the time complexity of Step(B) is indeed equal to that of Step(A). The same argument applies also to the function \tilde{F}_{lower} . Finally, it is also clear that the memory requirement of B is equal to the memory requirement of A. This completes the proof.