

Tight Security Bounds for Key-Alternating Ciphers

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Abstract. A t -round *key-alternating cipher* (also called *iterated Even-Mansour cipher*) can be viewed as an abstraction of AES. It defines a cipher E from t fixed public permutations $P_1, \dots, P_t : \{0, 1\}^n \rightarrow \{0, 1\}^n$ and a key $k = k_0 \| \dots \| k_t \in \{0, 1\}^{n(t+1)}$ by setting $E_k(x) = k_t \oplus P_t(k_{t-1} \oplus P_{t-1}(\dots k_1 \oplus P_1(k_0 \oplus x) \dots))$. The indistinguishability of E_k from a truly random permutation by an adversary who also has oracle access to the (public) random permutations P_1, \dots, P_t was investigated in 1997 by Even and Mansour for $t = 1$ and for higher values of t in a series of recent papers. For $t = 1$, Even and Mansour proved indistinguishability security up to $2^{n/2}$ queries, which is tight. Much later Bogdanov et al. (2011) conjectured that security should be $2^{\frac{t}{t+1}n}$ queries for general t , which matches an easy distinguishing attack (so security cannot be more). A number of partial results have been obtained supporting this conjecture, besides Even and Mansour's original result for $t = 1$: Bogdanov et al. proved security of $2^{\frac{2}{3}n}$ for $t \geq 2$, Steinberger (2012) proved security of $2^{\frac{3}{4}n}$ for $t \geq 3$, and Lampe, Patarin and Seurin (2012) proved security of $2^{\frac{t}{t+2}n}$ for all even values of t , thus “barely” falling short of the desired $2^{\frac{t}{t+1}n}$.

Our contribution in this work is to prove the long-sought-for security bound of $2^{\frac{t}{t+1}n}$, up to a constant multiplicative factor depending on t . Our method is essentially an application of Patarin's H-coefficient technique.

1 Introduction

Given t permutations $P_1, \dots, P_t : \{0, 1\}^n \rightarrow \{0, 1\}^n$ the t -round *key-alternating cipher* based on P_1, \dots, P_t is a blockcipher $E : \{0, 1\}^{(t+1)n} \times \{0, 1\}^n \rightarrow \{0, 1\}^n$ of keyspace $\{0, 1\}^{(t+1)n}$ and message space $\{0, 1\}^n$, where for a key $k = k_0 \| k_1 \| \dots \| k_t \in \{0, 1\}^{(t+1)n}$ and a message $x \in \{0, 1\}^n$ we set

$$E(k, x) = k_t \oplus P_t(k_{t-1} \oplus P_{t-1}(\dots P_1(k_0 \oplus x) \dots)). \quad (1)$$

(See Figure 1.) Plainly, $E(k, \cdot)$ is a permutation of $\{0, 1\}^n$ for each fixed $k \in \{0, 1\}^{(t+1)n}$; we let $E^{-1}(k, \cdot)$ denote the inverse permutation. The P_i 's are called the *round permutations* of E and t is the *number of rounds* of E . Thus t and the permutations P_1, \dots, P_t are parameters determining E .

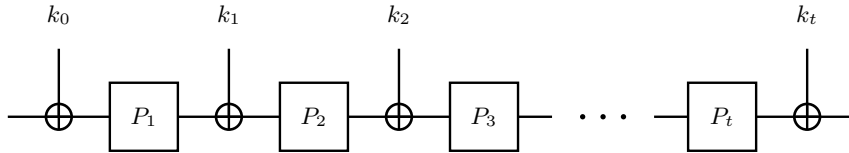


Fig. 1. A t -round key alternating cipher.

Key-alternating ciphers were first proposed (for values of t greater than 1) by the designers of AES [5, 6], the Advanced Encryption Standard. Indeed, AES-128 itself can be viewed as a particular

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instantiation of the key-alternating cipher paradigm in which the round permutations P_1, \dots, P_t equal a single permutation P (the Rijndael round function, in this case), in which $t = 10$, and in which only a subset of the $\{0, 1\}^{(t+1)n} = \{0, 1\}^{11n}$ possible keys are used (more precisely, the $11n$ bits of key are derived pseudorandomly from a seed of n bits, making the key space $\{0, 1\}^n = \{0, 1\}^{128}$). However, for $t = 1$ the design was proposed much earlier by Even and Mansour as a means of constructing a blockcipher from a fixed permutation [7]. Indeed, key-alternating ciphers also go by the name of *iterated Even-Mansour ciphers*.

Even and Mansour accompanied their proposal with “provable security” guarantees by showing that, for $t = 1$, an adversary needs roughly $2^{n/2}$ queries to distinguish $E(k, \cdot)$ for a random key k (k being hidden from the adversary) from a true random permutation, in a model where the adversary is given oracle access to $E(k, \cdot)$, $E^{-1}(k, \cdot)$ as well as to P_1 , P_1^{-1} , where P_1 is modeled as a random permutation (in the dummy world, the adversary is given oracle access to two independent random permutations and their inverses). Their bound was matched by Daemen [4], who showed a $2^{n/2}$ -query distinguishing attack for $t = 1$.

For $t > 1$, we can generalize the Even-Mansour indistinguishability experiment by giving the adversary oracle access to P_1, \dots, P_t and their inverses and to $E(k, \cdot)$, $E^{-1}(k, \cdot)$ in the real world (for a randomly chosen, hidden $k \in \{0, 1\}^{(t+1)n}$), and to a tuple of $t + 1$ independent random permutations and their inverses in the “ideal” or “dummy” world (see Figure 2). In this case, Daemen’s attack can be easily generalized to an attack of query complexity $2^{\frac{t}{t+1}n}$, as pointed out by Bogdanov et al. [2], but the security analysis of Even and Mansour could not be easily generalized to match this bound.

Bogdanov et al. did show, though, security of $2^{\frac{2}{3}n}$ for $t \geq 2$ (modulo lower-order terms), which is tight for $t = 2$ as it matches the $2^{\frac{2}{3}n}$ -query attack. Later Steinberger [19] improved this bound to $2^{\frac{3}{4}n}$ queries for $t \geq 3$ by modifying technical aspects of Bogdanov et al.’s analysis. Orthogonally and simultaneously, Lampe, Patarin and Seurin [13] used coupling-based techniques to show security of $2^{\frac{t}{t+1}n}$ queries for nonadaptive adversaries and security $2^{\frac{t}{t+2}n}$ for adaptive adversaries (and even values of t). While the bound $2^{\frac{t}{t+2}n}$ might seem “almost” sharp, we note that

$$2^{\frac{t}{t+2}n} = 2^{\frac{(t/2)}{(t/2)+1}n}$$

is actually the conjectured adaptive security for $t/2$ rounds. Indeed, Lampe et al. basically show that an adaptive adversary attacking the t -round construction has no more advantage than a nonadaptive adversary attacking $t/2$ rounds (this reduction follows upon work of Maurer et al. [16, 17]). Seen this way, Lampe et al.’s result appears less sharp. The issue is not only qualitative since their bound only improves on Steinberger’s for $t \geq 8$.

OUR RESULTS. In this paper we finally prove security of $2^{\frac{t}{t+1}n}$ queries for key-alternating ciphers, which has been the conjectured security since the paper of Bogdanov et al., and which is provably tight by the attack in the same paper. More precisely, we show that an adaptive adversary making at most q queries to each of its oracles has distinguishing advantage bounded by $O(1)q^{t+1}/N^t + O(1)$, where $N = 2^n$ and the two $O(1)$ terms depend on t . (See Section 2 for a formal statement.)

Our techniques are (maybe disappointingly) not as conceptually novel as those of [19] or [13], as we simply apply Patarin’s H-coefficient technique. The crucial step is lower bounding the probability of a certain event, namely of the event that q input-output values become linked when t partially defined composed permutations (whose composition so far poses no contradiction to the linking of said q input-output pairs) are randomly extended. The surprising aspect of these computations is that various “second-order” factors (that one might otherwise expect to not matter) actually need to be taken into account. Informally, this can be ascribed to the fact that the values of q under consideration are far beyond birthday.

Besides shedding some light on the structural and probabilistic aspects of key-alternating ciphers in the ideal permutation model, we also hope this paper will serve as a useful additional tutorial on (or introduction to) Patarin’s H-coefficient technique, which still seems to suffer from a lack of exposure.

We note that [13] also uses H-coefficient-based techniques and, indeed, our approach is much more closely inspired by that of [13] than by [2, 19].

PAPER ORGANIZATION. Definitions relating to key-alternating ciphers as well as a formal statement of our main result are given in Section 2. An overview of the H-coefficient technique is given in Section 3. The proof of the main theorem is given in Section 4, while a key lemma is proved in the paper’s full version [3].

EXTENSIONS. As we note in the proof, our main result holds even if the subkeys k_0, \dots, k_t are only t -wise independent instead of $(t + 1)$ -wise independent. This is particularly interesting for $t = 1$. Along different lines, and as pointed out to us by Jooyoung Lee, our result also implies tight security bounds for the “XOR-cascade” cipher introduced by Gaži and Tessaro [9, 10] via a reduction by Peter Gaži [10, 11].

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2 Definitions and Main Result

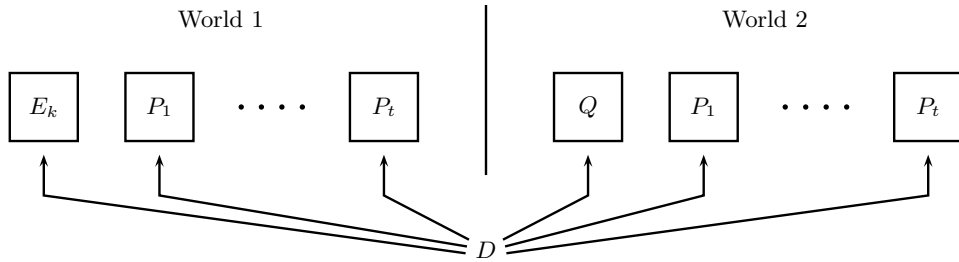


Fig. 2. The two worlds for the Even-Mansour security experiment. In World 1 the distinguisher D has oracle access to random permutations P_1, \dots, P_t and the key-alternating cipher E_k (cf. Eq. (1)) for a random key k . In World 2, D has oracle access to $t + 1$ independent random permutations. In either world D also has oracle access to the inverse of each permutation.

A t -round key-alternating cipher E has keyspace $\{0, 1\}^{(t+1)n}$ and message space $\{0, 1\}^n$. We refer back to equation (1) for the definition of $E(k, x)$ (which implicitly depends on the choice of round permutations P_1, \dots, P_t). We note that $E^{-1}(k, y)$ has an analogous formula in which $P_t^{-1}, \dots, P_1^{-1}$ are called. We write E_k for the permutation $E(k, \cdot)$.

We work in the ideal permutation model. For our purposes, the PRP security of a t -round key-alternating cipher E against a distinguisher (or “adversary”) D is defined as

$$\mathbf{Adv}_{E,t}^{\text{PRP}}(D) = \Pr[k = k_0 \cdots k_t \leftarrow \{0, 1\}^{(t+1)n}; D^{E_k, P_1, \dots, P_t} = 1] - \Pr[D^{Q, P_1, \dots, P_t} = 1] \quad (2)$$

where in each experiment Q, P_1, \dots, P_t are independent uniform random permutations, where D^A denotes that D has oracle access to A and A^{-1} (since all oracles are permutations), and where $k = k_0 \cdots k_t$ is selected uniformly at random (and hidden from D). See Figure 2. We further define

$$\mathbf{Adv}_{E,t}^{\text{PRP}}(q_e, q) = \max_D \mathbf{Adv}_{E,t}^{\text{PRP}}(D)$$

where the maximum is taken over all distinguishers D that make at most q_e queries to their first oracle and at most q queries to each of their other oracles. (The notation $\mathbf{Adv}_{E,t}^{\text{PRP}}(\cdot)$ is thus overloaded.)

Accounting for cipher queries and permutation queries separately has the main advantage of clarifying “which q is which” in the security bound. We also note that, besides t , n is a parameter on which E (and hence $\text{Adv}_{E,t}^{\text{PRP}}(q)$) depends.

(As an aside, we note the above indistinguishability experiment differs from the recently popular framework of *indifferentiability* by, among others, the presence of a secret key and the absence of a simulator; the similarity, on the other hand, is that the adversary can query the internal components of the structure. The end goal of the security proof is also different, since we simply prove PRP-security (with tight bounds) whereas indifferentiability aims to prove something much stronger, but, typically, with much inferior bounds. See [1, 14] for indifferentiability results on key-alternating ciphers.)

Our main result is the following:

Theorem 1. *Let $N = 2^n$ and let $q \leq N/3$, $t \geq 1$. Then for any constant $C > 0$,*

$$\text{Adv}_{E,t}^{\text{PRP}}(q_e, q) \leq \frac{q_e q^t}{N^t} \cdot C t^2 (6C)^t + (t+1)^2 \frac{1}{C}.$$

The presence of the adjustable constant C in Theorem 1 is typical of security proofs that involve a threshold-based “bad event”. The constant corresponds to the bad event’s (adjustable) threshold. Some terms in the security bound grow with C , others decrease with C , and for every q_e , q , t and N there is an optimal C . Choosing

$$C = \left(\frac{(t+1)N^t}{6^t t^2 q_e q^t} \right)^{1/(t+2)}$$

(which happens to be the analytical optimum) and using a little algebra yields the following, more readable corollary for the case $q = q_e$:

Corollary 1. *Let $N = 2^n$, $q \leq N/3$, $t \geq 1$. Then*

$$\text{Adv}_{E,t}^{\text{PRP}}(q, q) \leq (t+1)^2 (t+2) \left(\frac{6tq}{N^{t/(t+1)}} \right)^{(t+1)/(t+2)}. \quad (3)$$

Security therefore holds up to about $q \approx N^{\frac{t}{t+1}}/6t^4$, with “security exponent” $(t+1)/(t+2)$. Since t is typically viewed as a constant the polynomial factor $6t^4$ is not bothersome from the asymptotic point of view even though, obviously, such a factor considerably waters down the security bound for concrete parameters like $t = 10$, $n = 128$. We also note that if we fix q and N and let $t \rightarrow \infty$ then (3) becomes worse and worse (i.e., closer to 1 and eventually greater than 1) for sufficiently large t . This apparent security degradation is obviously an artefact of our bound, since a straightforward reduction shows that security can only increase with t .

3 The H-coefficient Technique in a Nutshell

In this section we give a quick high-level outline of Patarin’s H-coefficient technique. This tutorial takes a broader view than Patarin’s own [18], but [18] mentions refinements for nonadaptive adversaries and “plaintext only” attacks that we don’t touch upon here. We emphasize that the material in this section is “informal by design”.

The general setting is that of a q -query information-theoretic distinguisher D interacting with one of two oracles, the “real world” oracle or the “ideal world” oracle. (Each oracle might consist of several interfaces for D to query.) By such interaction, D creates a transcript, which is a list of queries made and answers returned. We can assume without loss of generality¹ that D is deterministic, and makes its final decision as a (deterministic) function of the transcript obtained.

¹ See Appendix A.

Denoting X the probability distribution on transcripts induced by the real world and denoting Y the probability distribution on transcripts induced by the ideal world (for some fixed deterministic distinguisher D) then D 's distinguishing advantage (cf. (2)) is easily seen to be upper bounded by

$$\Delta(X, Y) := \frac{1}{2} \sum_{\tau \in \mathcal{T}} |\Pr[X = \tau] - \Pr[Y = \tau]|$$

(the so-called *statistical distance* or *total variation distance* between X and Y) where \mathcal{T} denotes the set of possible transcripts.

The technique's central idea is to use the fact that

$$\Delta(X, Y) = 1 - E_{\tau \sim Y} [\min(1, \Pr[X = \tau] / \Pr[Y = \tau])] \quad (4)$$

in order to upper bound $\Delta(X, Y)$. Here $E_{\tau \sim Y} [Z(\tau)]$ is the expectation of the random variable $Z(\tau)$ when τ is sampled according to Y , and one assumes $\min(1, \Pr[X = \tau] / \Pr[Y = \tau]) = 1$ if $\Pr[Y = \tau] = 0$. For completeness we record the easy proof of (4):

$$\begin{aligned} \Delta(X, Y) &= \sum_{\tau \in \mathcal{T}: \Pr[Y = \tau] > \Pr[X = \tau]} (\Pr[Y = \tau] - \Pr[X = \tau]) \\ &= \sum_{\tau \in \mathcal{T}: \Pr[Y = \tau] > \Pr[X = \tau]} \Pr[Y = \tau] (1 - \Pr[X = \tau] / \Pr[Y = \tau]) \\ &= \sum_{\tau \in \mathcal{T}} \Pr[Y = \tau] (1 - \min(1, \Pr[X = \tau] / \Pr[Y = \tau])) \\ &= 1 - E_{\tau \sim Y} [\min(1, \Pr[X = \tau] / \Pr[Y = \tau])]. \end{aligned}$$

Thus, by (4), upper bounding the distinguisher's advantage reduces to lower bounding the expectation

$$E_{\tau \sim Y} [\min(1, \Pr[X = \tau] / \Pr[Y = \tau])]. \quad (5)$$

Typically, some transcripts are better than others, in the sense that for some transcripts τ the ratio

$$\Pr[X = \tau] / \Pr[Y = \tau]$$

might be quite small (when we would rather the ratio be near 1), but these "bad" transcripts occur with small probability. A typical proof classifies the set \mathcal{T} of possible transcripts into a finite number of combinatorially distinct classes $\mathcal{T}_1, \dots, \mathcal{T}_k$ and exhibits values $\varepsilon_1, \dots, \varepsilon_k \geq 0$ such that

$$\tau \in \mathcal{T}_i \implies \Pr[X = \tau] / \Pr[Y = \tau] \geq 1 - \varepsilon_i. \quad (6)$$

Then

$$E_{\tau \sim Y} [\min(1, \Pr[X = \tau] / \Pr[Y = \tau])] \geq \sum_{i=1}^k \Pr[Y \in \mathcal{T}_i] (1 - \varepsilon_i)$$

and, by (4),

$$\Delta(X, Y) \leq \sum_{i=1}^k \Pr[Y \in \mathcal{T}_i] \varepsilon_i.$$

The "ideal world" random variable Y often has a very simple distribution, making the probabilities $\Pr[Y \in \mathcal{T}_i]$ easy to compute. On the other hand, proving the lower bounds (6) for $i = 1 \dots k$ can be difficult, and we rediscuss this issue below.

Many proofs (including ours) have $k = 2$, with \mathcal{T}_1 consisting of the set of “good” transcripts and \mathcal{T}_2 consisting of the set of “bad” transcripts (i.e., those with small value of $\Pr[X = \tau]/\Pr[Y = \tau]$); then ε_1 is small and ε_2 is large, while (hopefully) $\Pr[Y \in \mathcal{T}_1]$ is large and $\Pr[Y \in \mathcal{T}_2]$ is small, and

$$\Delta(X, Y) \leq \Pr[Y \in \mathcal{T}_1]\varepsilon_1 + \Pr[Y \in \mathcal{T}_2]\varepsilon_2 \leq \varepsilon_1 + \Pr[Y \in \mathcal{T}_2].$$

The final upper bound on $\Delta(X, Y)$, in this case, can thus be verbalized as “one minus the probability ratio of good transcripts [i.e., ε_1], plus the probability of a transcript being bad” (the latter probability being computed with respect to the distribution Y). This is the form taken by our own bound.

LOWER BOUNDING THE RATIO $\Pr[X = \tau]/\Pr[Y = \tau]$. The random variables X and Y are, formally, defined on underlying probability spaces that contain respectively all the coins needed for the real and ideal world experiments. To be more illustrative, in the case of the key-alternating cipher distinguishability experiment X ’s underlying probability space consists of all possible $(t + 1)$ -tuples of the form (k, P_1, \dots, P_t) where $k \in \{0, 1\}^{(t+1)n}$ and where each P_i is a permutation of $\{0, 1\}^n$, while Y ’s underlying probability space is all $(t + 1)$ -tuples of the form (Q, P_1, \dots, P_t) where Q as well as each P_i is a permutation of $\{0, 1\}^n$. (In either case the measure is uniform, and for simplicity we also assume uniform—and hence finite—probability spaces in our discussion here.) For the following, we write Ω_X , Ω_Y for the probability spaces on which respectively X and Y are defined. We note that each ω in Ω_X or Ω_Y can be viewed as an oracle for D to interact with, thus we may use phrases such as “ D runs with oracle ω ”, etc. To summarize, X and Y are, formally, functions $X : \Omega_X \rightarrow \mathcal{T}$, $Y : \Omega_Y \rightarrow \mathcal{T}$, where $X(\omega)$ is the transcript obtained by running D with oracle $\omega \in \Omega_X$, and where $Y(\omega)$ is the transcript obtained by running D oracle $\omega \in \Omega_Y$.

There is usually an obvious notion of “compatibility” between a transcript τ and an element $\omega \in \Omega_X$ or $\omega \in \Omega_Y$. For example, in the case of key-alternating ciphers, if τ contains a query to P_1 and nothing else, the ω ’s in Ω_X that are compatible with τ will be exactly those where the P_1 -coordinate of ω agrees with the query in τ ; there are $2^{(t+1)n} \cdot (2^n - 1)! \cdot (2^n!)^{t-1}$ such “compatible” ω ’s in Ω_X . For the same transcript, there would be $(2^n - 1)! \cdot (2^n!)^t$ compatible ω ’s in Ω_Y . We write $\text{comp}_X(\tau)$ for the set of ω ’s in Ω_X compatible with a transcript τ , and we define $\text{comp}_Y(\tau)$ likewise with respect to Ω_Y .

We note that the statement “ ω is compatible with τ ” is actually not equivalent to the statement “running D with oracle ω produces τ ”. Indeed, some τ ’s may never be produced by D at all; e.g., if a transcript τ contains more than q queries, or if it contains queries to P_1 when D is a distinguisher that never queries P_1 , etc, then τ is never produced by D (i.e., $\Pr[X = \tau] = \Pr[Y = \tau] = 0$), but this does not prevent $\text{comp}_X(\tau)$, $\text{comp}_Y(\tau)$ from being well-defined.

A central insight of the H-coefficient technique (but which is usually taken for granted and used without mention) is that when τ is a possible transcript of D at all (i.e., if either $\Pr[X = \tau] > 0$ or $\Pr[Y = \tau] > 0$) then

$$\Pr[X = \tau] = \frac{|\text{comp}_X(\tau)|}{|\Omega_X|} \quad \text{and} \quad \Pr[Y = \tau] = \frac{|\text{comp}_Y(\tau)|}{|\Omega_Y|}. \quad (7)$$

These equalities, argued below, might seem obvious (or not) but one should note they carry some counterintuitive consequences. Firstly:

(c1) *The order in which queries appear in a transcript τ does not affect the probability of τ occurring; only the set of queries appearing in τ matters.*

(This because the sets $\text{comp}_X(\tau)$, $\text{comp}_Y(\tau)$ are unaffected by the order with which queries appear in τ .) Along the same lines, one has:

(c2) *If two different (deterministic) distinguishers can obtain a transcript τ each with nonzero probability, these distinguishers will obtain τ with equal probability. Moreover, by (c1), this holds even if the transcript carries no information about the order in which queries are made.*

(This because the right-hand sides in (7) are distinguisher-independent.) Thus, if D_1 and D_2 are two adaptive, deterministic distinguishers that can arrive (by a potentially completely different query order) at transcripts τ_1 and τ_2 that contain the same *set* of queries, then D_1 has the same probability of obtaining τ_1 as D_2 has of obtaining τ_2 , with this equality holding separately both in the real and ideal worlds. While very basic, the order-independence property (c1) and distinguisher-independence property (c2) of deterministic distinguishers seem not to have been highlighted anywhere before².

We now informally argue (7), focusing on the first equality (the X -world) for concreteness. Firstly, executing D with an $\omega \in \Omega_X$, $\omega \notin \text{comp}_X(\tau)$ can obviously not produce τ as a transcript, since ω is not compatible with τ . It therefore suffices to show that running D on an oracle $\omega \in \text{comp}_X(\tau)$ produces the transcript τ . For this, we know by assumption that there exists³ an $\omega' \in \Omega_X \cup \Omega_Y$ such that running D on oracle ω' produces τ . However, one can show by induction on the number of queries made by D that the computations D^ω and $D^{\omega'}$ will not “diverge”, since every time D makes a query to ω' this query appears in τ and, hence, because $\omega \in \text{comp}_X(\tau)$, will be answered the same by ω (also recall that D is deterministic). Hence D^ω will produce the same transcript as $D^{\omega'}$, i.e., τ .

By (7), the ratio $\Pr[X = \tau]/\Pr[Y = \tau]$ is equal to

$$\frac{\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]}{\Pr_{\Omega_Y}[\omega \in \text{comp}_Y(\tau)]}. \quad (8)$$

Here $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)] = |\text{comp}_X(\tau)|/|\Omega_X|$, $\Pr_{\Omega_Y}[\omega \in \text{comp}_Y(\tau)] = |\text{comp}_Y(\tau)|/|\Omega_Y|$ are different notations⁴ for the ratios appearing in (7).

Looking at (8) it is possible to wonder whether anything substantial has been gained so far, or whether notations are simply being shuffled around; after all, $\Pr[X = \tau]$ and $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]$ are “obviously the same thing”⁵ (and the same for Y). However the probability $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]$ offers a considerable conceptual advantage over the probability $\Pr[X = \tau]$, as $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]$ refers to an experiment with a non-adaptive flavor (a transcript τ is fixed, and a uniform random element of Ω_X is drawn—what is the probability of compatibility?) while the probability $\Pr[X = \tau]$ refers, by definition, to the adaptive interaction of D with its oracle, which is much messier to think about. Indeed, (c1) and (c2) already show that adaptivity is in a sense “thrown out” when (7) is applied.

4 Proof of Theorem 1

We make the standard simplifying assumption that the distinguisher D is deterministic. For simplicity, moreover, we assume that D makes exactly q_e queries to its first oracle and exactly q queries to each of its other oracles. This is without loss of generality.

We refer to the case where D has an oracle tuple of the type (E_k, P_1, \dots, P_t) as the “real world” and to the case when D has an oracle tuple of the type (Q, P_1, \dots, P_t) as the “ideal world”. For convenience, we will be generous with the distinguisher in the following way: at the end of the experiment (when the distinguisher has made its $(t + 1)q$ queries, but before the distinguisher outputs its decision) we reveal the key $k = k_0 k_1 \dots k_t$ to the distinguisher in the real world, while in the ideal world we sample a dummy key $k' = k'_0 k'_1 \dots k'_t$ and reveal this dummy key to the distinguisher. A distinguisher playing this “enhanced” game is obviously at no disadvantage, since it can disregard the key if it wants.

² A bit of thought reveals that (c1), (c2) hold for any experiment involving *stateless* oracles. More precisely, the oracle’s answer is a deterministic function of a random tape sampled at the beginning of the experiment.

³ Here ω' could also lie outside $\Omega_X \cup \Omega_Y$; the argument goes through as long as there exists *some* oracle leading to the transcript τ .

⁴ In fact, replacing $|\text{comp}_X(\tau)|/|\Omega_X|$ and $|\text{comp}_Y(\tau)|/|\Omega_Y|$ by respectively $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]$ and $\Pr_{\Omega_Y}[\omega \in \text{comp}_Y(\tau)]$ in (7) gives a more general formulation of these identities, for cases where the probability distributions on Ω_X , Ω_Y are not uniform. We prefer the fractions $|\text{comp}_X(\tau)|/|\Omega_X|$, $|\text{comp}_Y(\tau)|/|\Omega_Y|$ because these expressions seem more concrete.

⁵ In fact, as already pointed out, $\Pr[X = \tau]$ and $\Pr_{\Omega_X}[\omega \in \text{comp}_X(\tau)]$ are *not* the same thing for τ ’s outside the range of D .

For the remainder of the proof we consider a fixed distinguisher D conforming to the conventions above. We can summarize D 's interaction with its oracles by a transcript consisting of a sequence of tuples of the form (i, σ, x, y) where $i \in \{0, \dots, t\}$, $\sigma \in \{+, -\}$ and $x, y \in \{0, 1\}^n$, plus the key value k at the end of the transcript. If $\sigma = +$ such a tuple denotes that D made the query $P_i(x)$ obtaining answer y , or if $\sigma = -$ that D made the query $P_i^{-1}(y)$ obtaining answer x , and D 's interaction with its oracles (as well as D 's final output bit) can be uniquely reconstructed from such a sequence of tuples. In fact, we can (and shall) encode the transcript as an *unordered set of directionless* tuples of the form (i, x, y) (plus the key value k). Indeed, given that D is deterministic, D 's interaction can still be reconstructed from such a transcript. (Consider that D always makes the same first query, since it is deterministic; we can look up the answer to this query in the transcript, deduce the second query made by D again since D is deterministic, and so on.) All in all, therefore, the transcript can be encoded as a tuple $(k, p_0, p_1, \dots, p_t)$ where $k \in \{0, 1\}^{(t+1)n}$ is the key (real or dummy) and where p_i , $i \geq 1$, is a table containing q pairs (x, y) , where each such pair either indicates a query $P_i(x) = y$ or a query $P_i^{-1}(y) = x$ (which it is can be deduced from the transcript), and where p_0 similarly contains the q_e input-output pairs queried to the cipher. One can also view p_i as a bipartite graph with shores $\{0, 1\}^n$ and containing q (resp. q_e , in the case of p_0) disjoint edges.

We let \mathcal{T} denote the set of all possible transcripts, i.e., the set of all tuples of the form (k, p_0, \dots, p_t) as described above. We note that some elements of \mathcal{T} —in fact, most elements—may never be obtained by D . For example, if D 's first query is $P_1(0^n)$ then (this first query never varies and) any transcript obtained by D contains a pair of the form $(0^n, y)$ in the table p_1 , for some $y \in \{0, 1\}^n$.

Let \mathcal{P} be the set of all permutations of $\{0, 1\}^n$; thus $|\mathcal{P}| = (2^n)!$. Let $\mathcal{P}^t = \mathcal{P} \times \dots \times \mathcal{P}$ be the t -fold direct product of \mathcal{P} . Let $\Omega_X = \{0, 1\}^{(t+1)n} \times \mathcal{P}^t$ and let $\Omega_Y = \{0, 1\}^{(t+1)n} \times \mathcal{P}^{t+1}$. In the obvious way, elements of Ω_X can be viewed as real world oracles for D while elements of Ω_Y can be viewed as “ideal world” oracles for D . (We note that Ω_Y is slightly different from the Ω_Y appearing in the discussion of Section 3, due to our convention of giving away the key as part of the transcript.) We write $X(\omega)$ for the transcript obtained by running D with oracle $\omega \in \Omega_X$, and $Y(\omega)$ for the transcript obtained by running D with oracle $\omega \in \Omega_Y$. By endowing Ω_X , Ω_Y with the uniform probability distribution, X and Y become random variables of range \mathcal{T} , whose distributions are exactly those obtained by running D in the real and ideal worlds respectively.

Since D 's output is a deterministic function of the transcript, D 's distinguishing advantage is upper bounded by $\Delta(X, Y)$. In order to upper bound $\Delta(X, Y)$ we make use of the equality

$$\Delta(X, Y) = 1 - E_{\tau \sim Y} [\min(1, \Pr[X = \tau] / \Pr[Y = \tau])]$$

mentioned in Section 3. More precisely, we will identify a set $\mathcal{T}_1 \subseteq \mathcal{T}$ of “good” query transcripts, and a set $\mathcal{T}_2 \subseteq \mathcal{T}$ of “bad” transcripts, such that \mathcal{T} is the disjoint union of \mathcal{T}_1 and \mathcal{T}_2 . Then, as shown in Section 3,

$$\Delta(X, Y) \leq \varepsilon_1 + \Pr[Y \in \mathcal{T}_2] \tag{9}$$

where ε_1 is a number such that

$$\frac{\Pr[X = \tau]}{\Pr[Y = \tau]} \geq 1 - \varepsilon_1$$

for all $\tau \in \mathcal{T}_1$ such that $\Pr[Y = \tau] > 0$. Theorem 1 will follow by showing that

$$\Pr[Y \in \mathcal{T}_2] \leq (t+1)^2 \frac{1}{C} \quad \text{and} \quad \tau \in \mathcal{T}_1 \implies \frac{\Pr[X = \tau]}{\Pr[Y = \tau]} \geq 1 - \varepsilon_1 \tag{10}$$

where C is a constant appearing in the definition of a “bad” transcript, and where $\varepsilon_1 = q_e \left(\frac{q}{N}\right)^t C t^2 (6C)^t$ is the first term appearing in the bound of Theorem 1. For the remainder of the proof we assume that $C q_e q^t < N^t$. This is without loss of generality since Theorem 1 is vacuous otherwise.

BAD TRANSCRIPTS. Let $\tau = (k, p_0, p_1, \dots, p_t) \in \mathcal{T}$ be a transcript. We associate to τ a graph $G(\tau)$, dubbed the *round graph*, that encodes the information contained in k as well as in p_1, \dots, p_t (but that ignores p_0). $G(\tau)$ has $2(t+1) \cdot 2^n$ vertices, grouped into “shores” of size 2^n each, with each shore being identified with a copy $\{0, 1\}^n$. We index the $2(t+1)$ shores as $0^-, 0^+, 1^-, 1^+, \dots, t^-, t^+$. Vertex y in shore i^- is connected to vertex $y \oplus k_i$ in shore i^+ by an edge, and these are the only edges between shores i^- and i^+ . Moreover, for each $(x, y) \in p_i$, $1 \leq i \leq t$, we connect vertex x in shore $(i-1)^+$ to vertex y in shore i^- . Thus $G(\tau)$ consists of $(t+1)$ full bipartite matchings (one per subkey) alternately glued with q -edge partial matchings (one for each p_i , $1 \leq i \leq t$). Since $G(\tau)$ encodes all the information in k, p_1, \dots, p_t , we can also write a transcript τ in the form $\tau = (p_0, G)$ where $G = G(\tau)$.

Obviously, the presence of the full bipartite graphs corresponding to the subkeys k_0, \dots, k_t within $G(\tau)$ is not topologically interesting. Call an edge of $G(\tau)$ a “key edge” if the edge joins the shores i^-, i^+ for some $i \in \{0, \dots, t\}$. We then define the *contracted round graph* $\tilde{G}(\tau)$ obtained from $G(\tau)$ by contracting all key edges; thus $\tilde{G}(\tau)$ has only $t+1$ shores; moreover, when an edge $(y, y \oplus k_i)$ between shores i^-, i^+ of $G(\tau)$ is contracted, the resulting vertex of $\tilde{G}(\tau)$ is given label y if $0 \leq i \leq t-1$, and is given label $y \oplus k_i$ if $i = t$. (The labeling of vertices of $\tilde{G}(\tau)$ is somewhat unimportant and arbitrary, but we adopt the above convention so that vertices in shores 0^- and t^+ of $G(\tau)$ keep their original labels in $\tilde{G}(\tau)$. The latter ensures compatibility between these vertex labels and triples in p_0 .) We note that a transcript τ is not determined by the pair $(p_0, \tilde{G}(\tau))$ (the key material being unrecoverable from the latter pair) but, as we will see, $\Pr[X = \tau]$ is determined by $(p_0, \tilde{G}(\tau))$.

An edge between shores $(i-1)$ and i of $\tilde{G}(\tau)$ is called an *i-edge*. (Each *i-edge* arises from an entry in p_i .) We write $Z_{ij}(\tilde{G}(\tau))$ for the set of (necessarily edge-disjoint) paths that exists between shores i and j of $\tilde{G}(\tau)$. We write $Z_{ij}^-(\tilde{G}(\tau)), Z_{ij}^+(\tilde{G}(\tau))$ for vertices of paths in $Z_{ij}(\tilde{G}(\tau))$ that are respectively in shores i and j of $\tilde{G}(\tau)$. We write $p_0^- = \{x : (x, y) \in p_0\}$ and $p_0^+ = \{y : (x, y) \in p_0\}$ be the projection of p_0 to its first and second coordinates respectively.

We say a transcript τ is *bad* if there exist $0 \leq i < j \leq t$ such that

$$|Z_{ij}(\tilde{G}(\tau))| > \frac{Cq^{j-i}}{N^{j-i-1}} \quad (11)$$

or if there exists $0 \leq i \leq j \leq t$ such that

$$|\{(x, y) \in p_0 : x \in Z_{0,i}^-(\tilde{G}(\tau)) \wedge y \in Z_{j,t}^+(\tilde{G}(\tau))\}| > \frac{Cq_e q^{i+t-j}}{N^{i+t-j}}. \quad (12)$$

To motivate this definition we note that q^{j-i}/N^{j-i-1} is exactly the expected number of paths from shore i to shore j in the ideal world, whereas, likewise, $q_e q^{i+t-j}/N^{i+t-j}$ is the expected number of paths from shore j to shore i that “wrap around” through an edge in p_0 (though such edges are not encoded in $\tilde{G}(\tau)$ and, hence, such “wrap around” paths don’t physically exist in $\tilde{G}(\tau)$). The set of bad transcripts is denoted \mathcal{T}_2 and we let $\mathcal{T}_1 = \mathcal{T} \setminus \mathcal{T}_2$. Transcripts in \mathcal{T}_1 are called *good*.

The easy, Markov-inequality-based proof that $\Pr[Y \in \mathcal{T}_2] \leq (t+1)^2 \frac{1}{C}$ can be found in this paper’s full version [3].

LOWER BOUNDING $\Pr[X = \tau]/\Pr[Y = \tau]$ FOR $\tau \in \mathcal{T}_1$. An element $\omega = (k, P_1, \dots, P_t) \in \Omega_x$ is *compatible* with a transcript $\tau = (k^*, p_0, \dots, p_t)$ if $k = k^*$, if $P_i(x) = y$ for every $(x, y) \in p_i$, $1 \leq i \leq t$, and if $E_k(x) = y$ for every $(x, y) \in p_0$, where E_k stands for the Even-Mansour cipher instantiated with permutations P_1, \dots, P_t (and key k). We write $\text{comp}_X(\tau)$ for the set of w ’s in Ω_X that are compatible with τ .

Analogously, an $w = (k, P_0, P_1, \dots, P_t) \in \Omega_Y$ is compatible with τ if the same conditions as above are respected, but replacing the constraint $E_k(x) = y$ with $P_0(x) = y$ for $(x, y) \in p_0$. We write $\text{comp}_Y(\tau)$ for the set of w ’s in Ω_Y that are compatible with τ .

We also say $\omega = (k, P_1, \dots, P_t)$ is *partially compatible* with $\tau = (k^*, p_0, p_1, \dots, p_t)$ if $k = k^*$ and if $P_i(x) = y$ for all $(x, y) \in p_i$, $1 \leq i \leq t$. (Thus, the requirement that p_0 agrees with E_k is dropped for

partial compatibility.) Likewise $\omega \in \Omega_Y$ is *partially compatible* with τ if (exactly as above) $k = k^*$ and $P_i(x) = y$ for all $(x, y) \in p_i$, $1 \leq i \leq t$. (Thus, the requirement that p_0 agrees with P_0 is dropped.) We write $\text{comp}'_X(\tau)$, $\text{comp}'_Y(\tau)$ for the set of ω 's in, respectively, Ω_X or Ω_Y that are partially compatible with τ . Note that

$$\frac{|\text{comp}'_X(\tau)|}{|\Omega_X|} = \frac{|\text{comp}'_Y(\tau)|}{|\Omega_Y|} = \frac{1}{N^{t+1}} \cdot \prod_{i=1}^t \frac{(N - |p_i|)!}{N!} \quad (13)$$

for any transcript $\tau = (k, p_0, p_1, \dots, p_t)$, where $|p_i|$ denotes the number of pairs in p_i .

We say that a transcript $\tau \in \mathcal{T}$ is *attainable* if $\Pr[Y = \tau] > 0$. (Note that $\Pr[X = \tau] > 0 \implies \Pr[Y = \tau] > 0$.) In other words, a transcript is attainable if there exists an $\omega \in \Omega_Y$ such that D^ω produces the transcript τ .

It is necessary and sufficient to lower bound $\Pr[X = \tau] / \Pr[Y = \tau]$ for attainable transcripts $\tau \in \mathcal{T}_1$. By (7) and (13),

$$\frac{\Pr[X = \tau]}{\Pr[Y = \tau]} = \frac{|\text{comp}_X(\tau)|}{|\text{comp}'_X(\tau)|} \bigg/ \frac{|\text{comp}_Y(\tau)|}{|\text{comp}'_Y(\tau)|} \quad (14)$$

for τ such that $\Pr[Y = \tau] > 0$. (We emphasize that both equalities in (7) hold as long as D produces τ as a transcript on *some* oracle in $\Omega_X \cup \Omega_Y$.) For the remainder of the argument we fix an arbitrary transcript $\tau = (k, p_0, p_1, \dots, p_t) \in \mathcal{T}_1$. We aim to lower bound the right-hand side fraction in (14).

For random permutations P_1, \dots, P_t and partial permutations p_1, \dots, p_t , let $P_i \downarrow p_i$ denote the event that P_i extends p_i , i.e., that $P_i(x) = y$ for all $(x, y) \in p_i$; then it is easy to see that

$$\frac{|\text{comp}_X(\tau)|}{|\text{comp}'_X(\tau)|} = \Pr [E_k \downarrow p_0 \mid k, P_1 \downarrow p_1, \dots, P_t \downarrow p_t] \quad (15)$$

where the underlying probability space is the choice of the uniform random permutations P_1, \dots, P_t (the notation conditions on τ 's key k only to emphasize that k is not randomly chosen) and where $E_k \downarrow p_0$ is the event that $E_k(x) = y$ for all $(x, y) \in p_0$, where E_k is the Even-Mansour cipher with key k and permutations P_1, \dots, P_t . Similarly,

$$\frac{|\text{comp}_Y(\tau)|}{|\text{comp}'_Y(\tau)|} = \Pr [P_0 \downarrow p_0 \mid k, P_1 \downarrow p_1, \dots, P_t \downarrow p_t]$$

where the underlying probability space is the uniform random choice of P_0, P_1, \dots, P_t . In the latter conditional probability, however, the event $P_0 \downarrow p_0$ is independent of the conditioned premise, so

$$\frac{|\text{comp}_Y(\tau)|}{|\text{comp}'_Y(\tau)|} = \Pr [P_0 \downarrow p_0] = \prod_{\ell=0}^{q_e-1} \frac{1}{N - \ell}. \quad (16)$$

To facilitate the computation of the conditional probability that appears in (15), let (in accordance with the definition of the graph $\tilde{G}(\tau)$ above) \tilde{p}_i be defined by

$$(x, y) \in \tilde{p}_i \iff (x \oplus k_{i-1}, y) \in p_i$$

for $1 \leq i \leq t-1$, and by

$$(x, y) \in \tilde{p}_t \iff (x \oplus k_{i-1}, y \oplus k_i) \in p_i$$

for $i = t$. Thus $\tilde{p}_1, \dots, \tilde{p}_t$ are the t edge sets of the graph $\tilde{G}(\tau)$, i.e., \tilde{p}_i is the set of edges between shores $i-1$ and i of $\tilde{G}(\tau)$. By elementary considerations, one has

$$\Pr [E_k \downarrow p_0 \mid k, P_1 \downarrow p_1, \dots, P_t \downarrow p_t] = \Pr [E_0 \downarrow p_0 \mid P_1 \downarrow \tilde{p}_1, \dots, P_t \downarrow \tilde{p}_t] \quad (17)$$

where E_0 denotes the Even-Mansour cipher instantiated with key $0^{(t+1)n}$, and where the probability is taken (on either side) over the choice of the uniform random permutations P_1, \dots, P_t . We will therefore focus on the right-hand side probability in (17).

We say shore i of $\tilde{G}(\tau)$ is “to the left” of shore j if $i < j$. We also view paths in $\tilde{G}(\tau)$ as oriented from left to right: the path “starts” at the leftmost vertex and “ends” at the rightmost vertex.

Let $(x_1, y_1), \dots, (x_{q_e}, y_{q_e})$ be the q_e edges in p_0 . We write $R(x_\ell)$ for the rightmost vertex in the path of $\tilde{G}(\tau)$ starting at x_ℓ , and $L(y_\ell)$ for the leftmost vertex in the path of $\tilde{G}(\tau)$ ending at y_ℓ . (More often than not, x_ℓ and y_ℓ are not adjacent to any edges of $\tilde{G}(\tau)$, in which case $R(x_\ell) = x_\ell$, $L(y_\ell) = y_\ell$.) We write the index of the shore containing vertex v as $\text{Sh}(v)$. (Thus $\text{Sh}(v) \in \{0, 1, \dots, t\}$.) Because τ is good, and because we are assuming $Cq_e(q/N)^t < 1$, $\text{Sh}(R(x_\ell)) < \text{Sh}(L(y_\ell))$ for $1 \leq \ell \leq q_e$.

A vertex in shore $i \geq 1$ is *left-free* if it is not adjacent to a vertex in shore $i - 1$. A vertex in shore $i \leq t - 1$ is *right-free* if it is not adjacent to a vertex in shore $i + 1$.

To compute the conditional probability

$$\Pr [E_0 \downarrow p_0 \mid P_1 \downarrow \tilde{p}_1, \dots, P_t \downarrow \tilde{p}_t]$$

we imagine the following experiment in q_e stages. Let $G_0 = \tilde{G}(\tau)$. At the ℓ -th stage, G_ℓ is inductively defined from $G_{\ell-1}$. Let \tilde{p}_i^ℓ be the edges between shore $i - 1$ and i of G_ℓ . Initially, $G_\ell = G_{\ell-1}$. Then, as long as $R(x_\ell)$ is not in shore t , a value y is chosen uniformly at random from the set of left-free vertices in shore $\text{Sh}(R(x_\ell)) + 1$, and the edge $(R(x_\ell), y)$ is added to $\tilde{p}_{\text{Sh}(R(x_\ell))+1}^\ell$. G_ℓ is the result obtained when $R(x_\ell)$ reaches shore t . Thus, G_ℓ has at most t more edges than $G_{\ell-1}$.

Since the permutations P_1, \dots, P_t are uniformly random and independently chosen, it is easy to see that

$$\Pr [E_0 \downarrow p_0 \mid P_1 \downarrow \tilde{p}_1, \dots, P_t \downarrow \tilde{p}_t] = \Pr [G_{q_e} \downarrow p_0]$$

for the random graph G_{q_e} defined in the process above, where the notation $G_{q_e} \downarrow p_0$ is a shorthand to indicate that vertices x_ℓ and y_ℓ are connected by a path in G_{q_e} for $1 \leq \ell \leq q_e$. Moreover, writing $x_\ell \rightarrow y_\ell$ for the event that x_ℓ and y_ℓ are connected by a path in G_ℓ (and thus in G_{q_e}), and writing $G_\ell \downarrow p_0$ for the event $x_j \rightarrow y_j$ for $1 \leq j \leq \ell$, we finally find

$$\frac{|\text{comp}_X(\tau)|}{|\text{comp}'_X(\tau)|} = \prod_{\ell=0}^{q_e-1} \Pr[x_{\ell+1} \rightarrow y_{\ell+1} \mid G_\ell \downarrow p_0]. \quad (18)$$

This formula should be compared with (16). Indeed, (16) and (18) imply that

$$\frac{|\text{comp}_X(\tau)|}{|\text{comp}'_X(\tau)|} \bigg/ \frac{|\text{comp}_Y(\tau)|}{|\text{comp}'_Y(\tau)|} = \prod_{\ell=0}^{q_e-1} \frac{\Pr[x_{\ell+1} \rightarrow y_{\ell+1} \mid G_\ell \downarrow p_0]}{1/(N - \ell)} \quad (19)$$

which suggests that to lower bound $\Pr[X = \tau] / \Pr[Y = \tau]$ one should compare $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} \mid G_\ell \downarrow p_0]$ and $1/(N - \ell)$. (More specifically, give a lower bound for the former that is not much less than the latter.)

SOME PRELIMINARY QUANTITATIVE INTUITION FOR (19). At this stage we “pause” the proof to give some quantitative intuition about the product that appears in (19). The lower bounding of this product, indeed, is the heart of our proof. While discussing intuition we will make the simplifying assumption that $\text{Sh}(R(x_\ell)) = 0$, $\text{Sh}(L(y_\ell)) = t$ for all $1 \leq \ell \leq q_e$ (which, as it turns out, still captures the most interesting features of the problem).

As a warm-up we can consider the case $t = 1$. In this case, firstly, the “simplifying assumption” $\text{Sh}(R(x_\ell)) = 0$, $\text{Sh}(L(y_\ell)) = 1$ actually holds with probability 1 for all $\tau \in \mathcal{T}_1$, by the second bad event in the definition of a bad transcript (i.e., (12)), and by our wlog assumption that

$$1 > Cq_e(q/N)^t = Cq_e q/N. \quad (20)$$

(In more detail, the right-hand side of (12) is $Cq_e q/N$ for $i = j = 0$ or $i = j = 1$. Thus, if there exists an $(x_\ell, y_\ell) \in p_0$ such that either $R(x_\ell) = 1$ or $L(y_\ell) = 0$, then $\tau \in \mathcal{T}_2$.) Next (still for $t = 1$) it can be directly observed that

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] = \frac{1}{N - q - \ell}$$

since $\tilde{p}_1 = \tilde{p}_1^0$ contains q edges and since ℓ additional edges have been drawn by the time $G_{\ell+1}$ is constructed. In fact, the ratio $1/(N - q - \ell)$ is *greater* than $1/(N - \ell)$, which means that in this case the product (19) is also greater than 1, and one can therefore use $\varepsilon_1 = 0$. I.e., for $t = 1$ the distinguisher's advantage is upper bounded by

$$\varepsilon_1 + \Pr[Y \in \mathcal{T}_2] \leq 0 + \Pr[Y \in \mathcal{T}_2] \leq \frac{2q_e q}{N}$$

where the last inequality is obtained by direct inspection of the event $\tau \in \mathcal{T}_2$ for $t = 1$. (For $t = 1$, the only thing that can cause a transcript to be bad is if $p_0^- \oplus k_0 \cap p_1^- \neq \emptyset$ or if $p_0^+ \oplus k_1 \cap p_1^+ \neq \emptyset$.) Note that even while $\Pr[X = \tau]/\Pr[Y = \tau] \geq 1$ for all $\tau \in \mathcal{T}_1$ such that $\Pr[Y = \tau] > 0$, one has $\Pr[X = \tau]/\Pr[Y \in \mathcal{T}_1] = 0$ for most $\tau \in \mathcal{T}_2$ such that $\Pr[Y = \tau] > 0$. This is why ε_1 can attain zero.

In passing, note we have proved the $(2q_e q/N)$ -security of the key-alternating cipher for $t = 1$, which exactly recovers Even and Mansour's original result for $t = 1$. The difference is that the H-coefficient technique "mechanizes" the bound-proving, to a certain extent. (Even and Mansour's proof [7] is more complicated, though it pursues the same basic idea. See also Kilian and Rogaway's paper on DESX [12] for a nice game-based take on this argument.)

Given these auspicious beginnings for $t = 1$ one might feel inclined to optimism and to conjecture, say, that the product (19) is *always* greater than 1 for good transcripts. However, these hopes are quickly dashed by the case $t = 2$. We do an example. For this example, assume that \tilde{p}_1 and \tilde{p}_2 are disjoint, i.e., no edge in \tilde{p}_1 touches an edge in \tilde{p}_2 . (Thus $G_0 = \tilde{G}(\tau)$ contains no paths of length 2.) The example will be clearer if we start by examining the case $\tilde{p}_1 = \emptyset$ (i.e., when there are *no* edges between shore 0 and shore 1). Then one can compute that⁶

$$\Pr[x_1 \rightarrow y_1] = \left(1 - \frac{|\tilde{p}_2|}{N}\right) \frac{1}{N - |\tilde{p}_2|} = \left(\frac{N - |\tilde{p}_2|}{N}\right) \frac{1}{N - |\tilde{p}_2|} = \frac{1}{N}$$

and more generally, one similarly computes

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] = \left(1 - \frac{|\tilde{p}_2|}{N - \ell}\right) \frac{1}{N - \ell - |\tilde{p}_2|} = \frac{1}{N - \ell}. \quad (21)$$

for all $0 \leq \ell \leq q_e - 1$, since the vertex sampled in shore 1 to which $x_{\ell+1}$ is connected is sampled uniformly from a set of size $N - \ell$, and similarly the new vertex sampled in shore 2 (if such vertex is sampled) comes uniformly from a set of size $N - \ell - |\tilde{p}_2|$. So far, so good: (21) is exactly the same probability as in the ideal case.

Now we remove the assumption $\tilde{p}_1 = \emptyset$, but keep the assumption that \tilde{p}_1 and \tilde{p}_2 are disjoint. In this case, one has

$$\Pr[x_1 \rightarrow y_1] = \left(1 - \frac{|\tilde{p}_2|}{N - |\tilde{p}_1|}\right) \frac{1}{N - |\tilde{p}_2|} = \left(\frac{N - 2q}{N - q}\right) \frac{1}{N - q} = \frac{N - 2q}{(N - q)^2}.$$

As our interest is to compare this quantity to $1/N$, we further massage this expression by writing

$$\frac{N - 2q}{(N - q)^2} = \frac{1}{N} - \frac{1}{N} + \frac{N - 2q}{(N - q)^2} = \frac{1}{N} - \frac{(N - q)^2}{N(N - q)^2} + \frac{N(N - 2q)}{N(N - q)^2} = \frac{1}{N} - \frac{q^2}{N(N - q)^2}.$$

⁶ In more detail: when we travel from x_1 to y_1 , the sampling process first chooses a random endpoint in shore 1 to attach x_1 to, and this endpoint has probability $|\tilde{p}_2|/N$ of "hitting" an edge in \tilde{p}_2 (in which case we have no hope of reaching y_1). If we don't hit an edge in \tilde{p}_2 , there is further chance $1/(N - |\tilde{p}_2|)$ that we reach y_1 , since the vertex in shore 2 is sampled uniformly at random from a set of size $N - |\tilde{p}_2|$.

More generally, one finds that

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] = \left(1 - \frac{|\tilde{p}_2|}{N - \ell - |\tilde{p}_1|}\right) \frac{1}{N - \ell - |\tilde{p}_2|} = \frac{1}{N - \ell} - \frac{q^2}{(N - \ell)(N - \ell - q)^2} \quad (22)$$

as can be seen by substituting N by $N - \ell$ everywhere in the first computation. Thus the probability $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ is now slightly *lower* than $1/(N - \ell)$, which rules out the optimistic conjecture above. As for the value of the product (19) one finds, by (22),

$$\prod_{\ell=0}^{q_e-1} \left(1 - \frac{q^2}{(N - \ell - q)^2}\right) \geq \left(1 - \frac{q^2}{(N - 2q)^2}\right)^{q_e} \geq 1 - \frac{q_e q^2}{(N - 2q)^2}.$$

This is acceptably close to 1 (i.e., taking $\varepsilon_1 = q_e q^2 / (N - 2q)^2$ is acceptably close to zero) as long as $q_e q^2 \ll N^2$. We are (coincidentally or not, since the assumption $q_e q^2 \ll N^2$ has already been used to upper bound $\Pr[\tau \in \mathcal{T}_2]$) “bumping into” the security bound for $t = 2$. Thus, the approach still works for $t = 2$, but this time the approach “barely” works!

In fact, the simplifying assumption that \tilde{p}_1 and \tilde{p}_2 are disjoint can easily be removed since, as is not hard to see, having \tilde{p}_1 and \tilde{p}_2 disjoint is actually the worst case possible⁷ for $t = 2$.

Moreover, the initial simplifying assumption that $R(x_\ell) = 0$, $L(y_\ell) = 2$ for all ℓ is also easy to remove for $t = 2$, because $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ actually increases to $1/(N - q - \ell)$ (cf. the case $t = 1$) when either⁸ $R(x_\ell) = 1$ or $L(y_\ell) = 1$. Thus, the above computations essentially prove security of $q_e q^2 / N^2$ for $t \geq 2$ (indeed, security is easily seen to “transfer upwards” from smaller to larger values of t), which is the main result of Bogdanov et al. [2]. The proof sketched above is arguably simpler than Bogdanov et al.’s, though. (Also, Bogdanov et al. seem to forget that if the only goal is to prove security of $q_e q^2 / N^2$ for $t \geq 2$ it suffices to restrict oneself to the case $t = 2$. Their general approach, however, can be pushed slightly further to cover the case $t = 3$, as shown by Steinberger [19].)

We now consider the case $t = 3$. Already, doing an exact probability computation for the conditional probability $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ (as done in (22) for $t = 2$) promises to be quite tedious for $t = 3$, so we can look at doing back-of-the-envelope estimates instead. The simplest estimate is to lower bound the probability of $x_{\ell+1}$ reaching $y_{\ell+1}$ by upper bounding the probability that the path being constructed meets a pre-existing edge in either shore 1 or shore 2, viz.,

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] \geq \left(1 - \frac{2q}{N - \ell - q}\right) \frac{1}{N - \ell - q} \quad (23)$$

where $2q/(N - \ell - q)$ is a (crude) upper bound on the probability that the path touches a pre-existing edge in either shore 1 or shore 2, and where $1/(N - \ell - q)$ is the probability of reaching $y_{\ell+1}$ if the path reaches a right-free vertex in shore 2. However, (23) is *worse* than (22), so we are heading at best for security of $\varepsilon_1 \approx q_e q^2 / N^2$ if we use this estimate. One can argue that $2q/(N - \ell - q)$ can be replaced by $q/(N - \ell - q)$ in (23) (because: if we hit an edge in \tilde{p}_2 that is not adjacent to an edge in \tilde{p}_3 this only helps us, and if we hit an edge in \tilde{p}_2 that is adjacent to an edge in \tilde{p}_3 this can be “billed” to the corresponding edge in \tilde{p}_3) but even so we are headed towards a security of $q_e q^2 / N^2$, by comparison with (22). In fact, we can reflect that any approach that doesn’t somehow seriously take into account the presence of three rounds is doomed to fail, because the computation for $t = 2$ is actually tight (cf. footnote 7), and thus cannot be tweaked to give security better than $q_e q^2 / N^2$.

⁷ On the other hand, we cannot count on \tilde{p}_1 and \tilde{p}_2 having some small intersection in order to possibly repair our optimistic conjecture. Indeed, the distinguisher could make sure that \tilde{p}_1 and \tilde{p}_2 are almost certainly disjoint. For example, the distinguisher could make q P_2 -queries with values that start with $n/3$ 0’s, and also make q P_1^{-1} -queries with values that start with $n/3$ 0’s. Then \tilde{p}_1 and \tilde{p}_2 are disjoint unless the first $n/3$ bits of the key are 0, which occurs with negligible probability.

⁸ Note that one always has $R(x_\ell) < L(y_\ell)$ by the definition of \mathcal{T}_2 and by the wlog assumption $Cq_e q^t < N^t$.

As it turns out, the “exact but tedious” probability computation that we shied from above does deliver a bound that implies the desired security of $q_e q^3/N^3$, even while back-of-the-envelope estimates indicate a security bound of $q_e q^2/N^2$. Intuitively, the gain that occurs is due to the fact that when the path hits an edge of \tilde{p}_2 not connected to an edge of \tilde{p}_3 —and at most $Cq^2/N \ll q$ edges in \tilde{p}_2 are adjacent to edges in \tilde{p}_3 , by definition of \mathcal{T}_2 —this is actually better than not hitting any edge at all in shore 1, because it *guarantees* we won’t hit an edge in \tilde{p}_3 . While this intuition is easy to see, it is somewhat harder to believe such a small “second-order” effect would make a crucial difference in the final security bound. Yet, this is exactly so. In fact, given the “completeness” of the H-coefficient method it makes sense to have faith that the exact probability computation (if doable) will deliver security $q_e q^3/N^3$. (Though in reality even this is not a given: by giving away the key at the end of each transcript we have been more generous to the adversary than those who devised the security conjecture of $q_e q^t/N^t$, so it’s possible to conceive that it’s the “key’s fault” if the security is (apparently) topping off at $q_e q^2/N^2$ (as opposed to the fault of our lossy estimates). Note that even if we have the correct intuition, and we believe it isn’t the “key’s fault” and that the approach is theoretically sound, we are still up against the problem of actually doing the computations in a such way that the desired security gain becomes apparent, and isn’t lost in a sea of fractions.)

Before proceeding with the exact-but-tedious computation for $t = 3$ it will be useful if we first estimate what kind of lower bound is actually needed for $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ in order to reach overall security $\approx q_e q^t/N^t$. Writing

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] = \frac{1}{N - \ell} + z_t$$

where z_t is an “error term” whose magnitude will determine ε_1 , we find that

$$\prod_{\ell=0}^{q_e-1} \frac{\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]}{1/(N - \ell)} = \prod_{\ell=0}^{q_e-1} (1 - (N - \ell)z_t) \geq (1 - N|z_t|)^{q_e} \geq 1 - Nq_e|z_t|.$$

Thus we will have $\varepsilon_1 \approx Nq_e|z_t|$ and so we need need $Nq_e|z_t| \ll 1$ in order for ε_1 to be small. Having

$$|z_t| = q^t/N^{t+1} \tag{24}$$

gives us precisely this under the assumption $q_e q^t/N^t \ll 1$.

DETAILS ON THE CASE $t = 3$. Let U_{ij} be the set of paths from shore i to shore j in $G(\tau)$, $0 \leq i < j \leq 3$, such that the vertex of the path in shore i is left-free (i.e., is the head of the path), but where the vertex in shore j may or may not be right-free. The U_{ij} ’s are therefore “half-open” paths. Note $|U_{ij}| \leq |Z_{ij}| \leq Cq^{j-1}/N^{j-i-1}$ by definition of \mathcal{T}_2 . For notational consistency with Lemma 1 below we rename \tilde{p}_i as E_i for $i = 1, 2, 3$. Thus $|E_i| = q$ and E_i is the set of edges between shores $(i - 1)$ and i of $\tilde{G}(\tau)$. Moreover, one can note that $E_i = \bigcup_{0 \leq j < i} U_{ji}$ for all i , with the latter being a disjoint union.

We start by computing $\Pr[x_1 \rightarrow y_1]$, from which the general case $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ will be easy to deduce. We view the underlying probability space as the selection of three vertices u_1, u_2 and u_3 from shores 1, 2 and 3 of $\tilde{G}(\tau)$ respectively, such that u_i is selected independently and uniformly at random from the set of left-free vertices in shore i . This defines a path $w_0 := x_1, w_1 := u_1, w_2, w_3$ where w_2 equals u_2 if u_1 is right-free and equals the other endpoint of the edge adjacent to u_1 otherwise, and where w_3 equals u_3 if w_2 is right-free, otherwise equals the vertex in shore 3 adjacent to w_2 . Then $\Pr[x_1 \rightarrow y_1]$ is equal to the probability that $w_3 = y_1$.

Since y_1 is left-free we have

$$w_3 = y_1 \iff (u_3 = y_1) \wedge \neg(w_1 \in U_{13} \vee w_2 \in U_{23}).$$

(The event $\neg(w_1 \in U_{13} \vee w_2 \in U_{23})$ coincides with the event that w_2 is right-free.) Note the event $u_3 = y_1$ is independent from the event $\neg(w_1 \in U_{13} \vee w_2 \in U_{23})$, and also that the events $w_1 \in U_{13}$ and

$w_2 \in U_{23}$ are disjoint. Moreover,

$$w_2 \in U_{23} \iff (u_2 \in U_{23}) \wedge \neg(w_1 \in U_{12})$$

since the vertices in shore 2 of U_{23} are left-free. By independence of u_1 and u_2 , thus,

$$\begin{aligned} \Pr[w_2 \in U_{23}] &= \Pr[u_2 \in U_{23}] \cdot (1 - \Pr[w_1 \in U_{12}]) \\ &= \frac{|U_{23}|}{N - |E_2|} \left(1 - \frac{|U_{12}|}{N - |E_1|} \right) \\ &= \frac{|U_{23}|}{N - |E_2|} - \frac{|U_{12}||U_{23}|}{(N - |E_1|)(N - |E_2|)}. \end{aligned}$$

Thus

$$\begin{aligned} \Pr[w_3 = y_1] &= \Pr[u_3 = y_1](1 - \Pr[w_1 \in U_{13}] - \Pr[w_2 \in U_{23}]) \\ &= \frac{1}{N - |E_3|} \left(1 - \frac{|U_{13}|}{N - |E_1|} - \frac{|U_{23}|}{N - |E_2|} + \frac{|U_{12}||U_{23}|}{(N - |E_1|)(N - |E_2|)} \right) \\ &= \frac{1}{N - |E_3|} - \frac{|U_{13}|}{(N - |E_1|)(N - |E_3|)} - \frac{|U_{23}|}{(N - |E_2|)(N - |E_3|)} \\ &\quad + \frac{|U_{12}||U_{23}|}{(N - |E_1|)(N - |E_2|)(N - |E_3|)}. \end{aligned}$$

(Note that none of the terms above are as small as $\approx q^3/N^4$ (cf. (24)), even with the approximation $\frac{1}{N-|E_i|} \approx \frac{1}{N}$, so none of the terms above can (yet) be folded into the error term.) Adding and subtracting the “ideal” probability $\frac{1}{N}$ to $\frac{1}{N-|E_3|}$ gives

$$\frac{1}{N} - \frac{1}{N} + \frac{1}{N - |E_3|} = \frac{1}{N} + \frac{|E_3|}{N(N - |E_3|)} = \frac{1}{N} + \frac{|U_{03}| + |U_{13}| + |U_{23}|}{N(N - |E_3|)}$$

(Here $\frac{|U_{03}|}{N(N-|E_3|)}$ is basically the same order of magnitude as q^3/N^4 , given that $|U_{03}| \leq |Z_{03}| \leq Cq^3/N^2$. So we can leave this term alone.) Next,

$$\frac{|U_{13}|}{N(N - |E_3|)} - \frac{|U_{13}|}{(N - |E_1|)(N - |E_3|)} = -\frac{|E_1||U_{13}|}{N(N - |E_1|)(N - |E_3|)} = -\frac{|U_{01}||U_{13}|}{N(N - |E_1|)(N - |E_3|)}$$

(same order of magnitude as q^3/N^4 , given that $|U_{13}| \leq Cq^2/N$), and

$$\begin{aligned} \frac{|U_{23}|}{N(N - |E_3|)} - \frac{|U_{23}|}{(N - |E_2|)(N - |E_3|)} &= -\frac{|E_2||U_{23}|}{N(N - |E_2|)(N - |E_3|)} \\ &= -\frac{|U_{02}||U_{23}|}{N(N - |E_2|)(N - |E_3|)} - \frac{|U_{12}||U_{23}|}{N(N - |E_2|)(N - |E_3|)} \end{aligned}$$

where only $\frac{|U_{02}||U_{13}|}{N(N-|E_2|)(N-|E_3|)}$ is small enough to fit inside the error term. But then, of course, we lastly compute that

$$\begin{aligned} &-\frac{|U_{12}||U_{23}|}{N(N - |E_2|)(N - |E_3|)} + \frac{|U_{12}||U_{23}|}{(N - |E_1|)(N - |E_2|)(N - |E_3|)} \\ &= \frac{|E_1||U_{12}||U_{23}|}{N(N - |E_1|)(N - |E_2|)(N - |E_3|)} \\ &= \frac{|U_{01}||U_{12}||U_{23}|}{N(N - |E_1|)(N - |E_2|)(N - |E_3|)} \end{aligned}$$

which is small enough to fit inside the error term. Collecting the leftovers after the various cancellations above, thus, we find

$$\begin{aligned} \Pr[w_3 = y_1] &= \frac{1}{N} + \frac{|U_{03}|}{N(N - |E_3|)} - \frac{|U_{01}||U_{13}|}{N(N - |E_1|)(N - |E_3|)} \\ &\quad - \frac{|U_{02}||U_{13}|}{N(N - |E_1|)(N - |E_3|)} + \frac{|U_{01}||U_{12}||U_{23}|}{N(N - |E_1|)(N - |E_2|)(N - |E_3|)} \end{aligned} \quad (25)$$

where all the terms except $\frac{1}{N}$ are “error-term small”. Moreover, when we compute $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ for $\ell \geq 1$ we can discard the ℓ completed paths from shore 0 to shore 3 linking the vertex pairs $(x_1, y_1), \dots, (x_\ell, y_\ell)$, and thus reduce to the case $\ell + 1 = 1$ with N replaced by $N - \ell$. I.e., the expression for $\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]$ will be identical to (25) except with N replaced by $N - \ell$ throughout.

From here the proof for $t = 3$ can be finished without many surprises. The crux of the proof is indeed the very simple idea of adding and subtracting $\frac{1}{N}$ from the probability, and of letting cancellations occur. This approach is purely algebraic. When we carry out the same process for an arbitrary value of t (see the proof of Lemma 1 in the full version of this paper [3]) we adopt a more combinatorial approach that recasts the algebraic manipulations as manipulations of events, which seems more satisfying because it gives the algebraic cancellations a concrete probabilistic interpretation. We note that doing so requires enlarging the probability space beyond its original confines. Indeed, for example, the original probability space for $t = 3$ has no event that occurs with probability $\frac{1}{N}$ even while factors of $\frac{1}{N}$ are ubiquitous in the final expression.

UPSHOT. The lemma below essentially generalizes the computation for $t = 3$ to arbitrary t . In this lemma U_{ij} stands for the set of paths from shore i to shore j of G_ℓ such that the vertex in shore i is left-free but where, as before, the vertex in shore j may or may not be right-free.

Lemma 1. *We have, under the notations described above,*

$$\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] = \frac{1}{N - \ell} - \frac{1}{N - \ell} \sum_{\sigma \in \mathfrak{S}_\ell} (-1)^{|\sigma|} \prod_{j=1}^{|\sigma|} \frac{|U_{i_j i_{j-1}}|}{N - |E_{i_j}|}$$

for each ℓ , $0 \leq \ell \leq q_e - 1$, where \mathfrak{S}_ℓ is the set of all sequences $\sigma = (i_0, \dots, i_s)$ with $R(x_{\ell+1}) = i_0 < \dots < i_s = L(y_{\ell+1})$, and where $|\sigma| = s$.

The proof of this lemma is given in the paper’s full version [3].

FINISHING THE PROOF OF THEOREM 1. We now apply Lemma 1 to lower bounding the product (19). For $1 \leq r \leq t$, let

$$\mathcal{L}_r = \{\ell : L(y_\ell) - R(x_\ell) = r\} \subseteq \{1, \dots, q_e\}$$

where (we recall) the elements of p_0 are $(x_1, y_1), \dots, (x_{q_e}, y_{q_e})$. By the definition of \mathcal{T}_2 , $\mathcal{L}_1, \dots, \mathcal{L}_t$ cover $\{1, \dots, q_e\}$ (i.e., there is no ℓ with $R(x_\ell) \geq L(y_\ell)$). Note that $|U_{ij}| \leq Cq^{j-i}/N^{j-i-1}$ (by the definition of \mathcal{T}_2) for $0 \leq i < j \leq t$, and $|E_i| \leq q$ for $1 \leq i \leq r$. Thus for $\ell + 1 \in \mathcal{L}_r$ we obtain, by Lemma 1,

$$\begin{aligned} \Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0] &= \frac{1}{N - \ell} - \frac{1}{N - \ell} \sum_{\sigma \in \mathfrak{S}_\ell} (-1)^{|\sigma|} \prod_{h=1}^{|\sigma|} \frac{|U_{i_{h-1} i_h}|}{N - \ell - |E_{i_h}|} \\ &\geq \frac{1}{N - \ell} - \frac{1}{N - \ell} \sum_{\sigma \in \mathfrak{S}_\ell} \prod_{h=1}^{|\sigma|} \frac{Cq^{i_h - i_{h-1}}/N^{i_h - i_{h-1} - 1}}{N - \ell - q} \\ &= \frac{1}{N - \ell} - \frac{1}{N - \ell} 2^{r-1} \left(\frac{q}{N}\right)^r \left(\frac{CN}{N - \ell - q}\right)^{|\sigma|} \end{aligned}$$

$$\begin{aligned}
&\geq \frac{1}{N-\ell} - \frac{1}{N-\ell} \left(\frac{2q}{N}\right)^r \left(\frac{CN}{N-2q}\right)^r \\
&\geq \frac{1}{N-\ell} - \frac{1}{N-\ell} \left(\frac{6Cq}{N}\right)^r.
\end{aligned}$$

Moreover $|\mathcal{L}_r| \leq t \cdot \frac{Cq_e q^{t-r}}{N^{t-r}}$ by the definition of \mathcal{T}_2 , so

$$\begin{aligned}
\prod_{\ell+1 \in \mathcal{L}_r} \frac{\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]}{1/(N-\ell)} &\geq \prod_{\ell+1 \in \mathcal{L}_r} \left(1 - \left(\frac{6Cq}{N}\right)^r\right) \\
&\geq 1 - \frac{Ctq_e q^{t-r}}{N^{t-r}} \left(\frac{6Cq}{N}\right)^r \\
&= 1 - \frac{Ctq_e q^t}{N^t} (6C)^r
\end{aligned}$$

Thus

$$\begin{aligned}
\prod_{\ell=0}^{q_e-1} \frac{\Pr[x_{\ell+1} \rightarrow y_{\ell+1} | G_\ell \downarrow p_0]}{1/(N-\ell)} &\geq 1 - \sum_{r=1}^t \frac{Ctq_e q^t}{N^t} (6C)^r \\
&\geq 1 - \frac{q_e q^t}{N^t} C t^2 (6C)^t.
\end{aligned}$$

This means

$$\frac{\Pr[X = \tau]}{\Pr[Y = \tau]} \geq 1 - \varepsilon_1$$

for $\varepsilon_1 = \frac{q_e q^t}{N^t} C t^2 (6C)^t$, for all $\tau \in \mathcal{T}_1$ such that $\Pr[Y = \tau] > 0$. Together with the fact that $\Pr[Y \in \mathcal{T}_2] \leq (t+1)^2 \frac{1}{C}$ this concludes the proof of Theorem 1 by (9).

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A Derandomizing an Information-Theoretic Distinguisher

The fact that an information-theoretic distinguisher can be derandomized is seldom proved, though admittedly simple. For a change and for the sake of completeness we include a proof here.

Let D be an information-theoretic distinguisher, which we view as a deterministic function taking an *oracle* input ω and a *random string* input r , and producing one bit of output. Formally D is a function

$$D : \Omega \times \mathcal{R} \rightarrow \{0, 1\}$$

where Ω is the set of possible oracles and where \mathcal{R} is the set of possible random strings. The fact that an “oracle” is an object for D to “interact” with according to certain rules doesn’t matter here. All that matters that D defines a deterministic function from $\Omega \times \mathcal{R}$ to $\{0, 1\}$.

Let r be an arbitrary random variable of range \mathcal{R} and let ω_X, ω_Y be two random variables of range Ω , where ω_X is distributed according to the distribution of real-world oracles and ω_Y is distributed according to the distribution of ideal-world oracles, and where r is independent from ω_X, ω_Y . By definition D ’s advantage (with respect to source of randomness r) is

$$\Delta_D := \Pr_{\omega_X, r}[D(\omega_X, r) = 1] - \Pr_{\omega_Y, r}[D(\omega_Y, r) = 1] \quad (26)$$

which can also be written

$$\Delta_D = \Delta(D(\omega_X, r), D(\omega_Y, r)) \quad (27)$$

where, on the right, we have the statistical distance of the random variables $D(\omega_X, r), D(\omega_Y, r)$ of range $\{0, 1\}$. Note that the right-hand side of (26) can be written

$$\mathbb{E}_r[\mathbb{E}_{\omega_X}[D(\omega_X, r)]] - \mathbb{E}_r[\mathbb{E}_{\omega_Y}[D(\omega_Y, r)]]$$

since D is $\{0, 1\}$ -valued, and where \mathbb{E} denotes expectation. By linearity of expectation, then,

$$\Delta_D = \mathbb{E}_r[\mathbb{E}_{\omega_X}[D(\omega_X, r)] - \mathbb{E}_{\omega_Y}[D(\omega_Y, r)]]$$

and so there must exist some $r_0 \in \mathcal{R}$ such that

$$\begin{aligned} \Delta_D &\leq \mathbb{E}_{\omega_X}[D(\omega_X, r_0)] - \mathbb{E}_{\omega_Y}[D(\omega_Y, r_0)] \\ &= \Pr_{\omega_X}[D(\omega_X, r_0) = 1] - \Pr_{\omega_Y}[D(\omega_Y, r_0) = 1] \end{aligned}$$

so that D ’s random string can be fixed to r_0 without harming D ’s advantage. (The fact that r is independent from ω_X, ω_Y is used to condition on $r = r_0$ without affecting the distribution of ω_X, ω_Y .) Alternatively, one can use (27) together with the more general fact that

$$\Delta(f(X, Z), f(Y, Z)) \leq \mathbb{E}_Z[\Delta(f(X, Z), f(Y, Z))] := \sum_z \Pr[Z = z] \Delta(f(X, z), f(Y, z)) \quad (28)$$

for any random variables X, Y, Z such that Z is independent from X and Y , for any function f . But to be complete (28) would require its own proof.