Efficient Optimization of Cut-offs in Quantum Repeater Chains

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Abstract—Quantum communication enables the implementation of tasks that are unachievable with classical resources. However, losses on the communication channel preclude the direct long-distance transmission of quantum information in many relevant scenarios. In principle quantum repeaters allow one to overcome losses. However, realistic hardware parameters make long-distance quantum communication a challenge in practice. For instance, in many protocols an entangled pair is generated that needs to wait in quantum memory until the generation of an additional pair. During this waiting time the first pair decoheres, impacting the quality of the final entanglement produced. At the cost of a lower rate, this effect can be mitigated by imposing a cut-off condition. For instance, a maximum storage time for entanglement after which it is discarded.

Cut-offs have been considered for entanglement generation in different contexts [6]–[17]. Notably, they play a key role for generating entanglement already in multi-pair experiments between adjacent nodes [8]. They also promise to be helpful in near-term quantum repeater experiments [9], [10], [14]. In the multi-repeater case, it is possible to obtain analytical expressions for the waiting time for general families of protocols [15], [16], though in general it appears challenging to extend those methods to characterize the quality of the states produced. Santra et al. [11] analytically optimized the distillable entanglement for a restricted class of quantum repeater schemes.

In this work, we first characterize the performance of a very general class of repeater schemes including cut-offs, probabilistic swapping, distillation and memory decoherence. We sidestep the challenge of analytical characterization by computing the probability distribution of the waiting time and fidelity of the first generated entangled pair between the repeater’s end nodes. For this, we improve the closed-form expressions by Brand et al. [18] to get faster algorithm runtimes and extend the expressions to repeater schemes which involve distillation and cut-offs. The runtime of the algorithm which evaluates these expressions is polynomial in the pre-specified size of the computed probability distribution’s support.

In the second part of the paper, we optimize the choices of the cut-off to maximize the secret-key rate. We study different cut-off strategies and find that the use of the optimal cut-off extends the parameter regime for which secret key can be generated and moreover significantly increases the secret-key rate for a large range of parameters.

I. INTRODUCTION

The realization of a quantum internet [1] will allow any two parties on earth to implement tasks that are impossible with its classical counterpart [2]. Quantum communication schemes rely on the transmission of quantum information, which in practice is precluded over long distances due to loss in the communication channel (usually glass fiber or free space). This problem can be overcome by dividing the distance between the sender and receiver of the quantum information into smaller segments, which are connected by intermediate nodes called quantum repeaters [3].

Most repeater schemes require quantum memories [4], [5]. Moreover, in many protocols an entangled pair is generated that needs to wait in a quantum memory until the generation of an additional pair. During this waiting time the first pair decoheres, reducing the quality of the final entanglement produced. At the cost of a lower rate, this effect can be mitigated by imposing a cut-off condition. For instance, a maximum storage time for entanglement after which it is discarded [6].

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explain how they can be composed into a repeater scheme that spans multiple nodes.

The first block GEN represents the generation of fresh entanglement between two adjacent nodes. We refer to those entangled pairs as an elementary link. The GEN block thus spans precisely two nodes, takes no input and outputs a single link.

The second and third blocks are entanglement swap (SWAP) and entanglement distillation (DIST). In the setting of two nodes A and B with a middle station M in between, an entanglement swap [19] takes two links A ↔ M and M ↔ B and outputs a single link A ↔ B. It spans at least three nodes. Next, entanglement distillation probabilistically transforms two low-quality links between the same pair of nodes to a new one with higher quality [20], [21]. The DIST block thus spans at least two nodes, takes two links as input and outputs a single link, where each link is shared by the same pair of nodes. Both SWAP and DIST consist of local operations including measurements and classical communication. They can succeed or fail and in case of failure, both input links are lost.

The last PROTOCOL-UNIT is CUT-OFF, which takes two links as input (not necessarily between the same nodes). It accepts or rejects the two input links depending on a success condition. In case of success, it leaves the two input links untouched and outputs them again. In case of failure, both input links are discarded. In this work, we study three different success conditions. In the first two, DIF-TIME-CUT-OFF and MAX-TIME-CUT-OFF, ‘success’ is declared if respectively the difference or the maximum of the input links’ production times does not exceed some pre-specified cut-off threshold. In the third strategy, FIDELITY-CUT-OFF, the input states are passed on only if they are both of sufficient quality. This success condition translates to a cut-off on the individual input states’ fidelity with a maximally-entangled state (see section II-B).

We now explain how the PROTOCOL-UNITS described above can be composed into a single repeater protocol spanning a chain of nodes. See Fig. 1(b) for an example. Each composite protocol on a chain of nodes starts with one or multiple GEN blocks between each pair of adjacent nodes for fresh elementary link generation. A protocol then consists of stacking instances of the other three PROTOCOL-UNITS in such a way that the output link(s) of one are used as input link(s) to the other. The only restriction on how the PROTOCOL-UNITS can be stacked is that both output links of CUT-OFF are used as inputs for one DIST or SWAP block. As a consequence of the stacking, any repeater protocol in the class we study has a tree structure (see also Fig. 1b). If a block at the root of a tree fails, then its input links are discarded and the GEN blocks at the tree’s leaves will restart.

We note that the class of repeater protocols described above includes, for instance, the well-known family of repeater schemes described by Briegel et al. [3], [22].

### II. Preliminaries

#### A. Class of repeater protocols considered

A quantum repeater chain connects two endpoints via several repeaters and aims to generate entanglement between the endpoints. In this section, we elaborate on the class of quantum repeater chain protocols we study in this work, which is an extension of the class studied in [18] with the addition of cut-offs. While doing so, we refer to both the endpoints and the repeater stations as nodes and to an entangled state between two nodes as a link.

The class of quantum repeater protocols studied in this work are composed of the following four building blocks or PROTOCOL-UNITS: elementary link generation (GEN), entanglement swap (SWAP), entanglement distillation (DIST) and cut-off (CUT-OFF). See Fig. 1(a). All of these processes can fail, but the involved nodes receive a success or failure message. That is, they are heralded. In what follows, we describe these four PROTOCOL-UNITS in more detail and subsequently
to the modelling in [18], except for the newly introduced 
CUT
-off unit. For each 
PROTOCOL-UNIT
, we describe the success condition as well as the quantum state that it outputs.

First, we model the fresh entanglement generation (GEN) using schemes which generate links in heralded attempts of duration $L_{\text{intermode}}/c$, where $L_{\text{intermode}}$ is the internode distance and $c$ is the speed of light in the used transmission medium, e.g. glass fiber [4]. We assume that each attempt is independent and succeeds with constant probability $0 < p_{\text{gen}} \leq 1$. For simplicity, we assume that the nodes are equally spaced with internode distance $L_0$, so that each attempt in elementary link generation takes duration $\Delta t_0 = L_0/c$, which will be the time unit in our numerical simulation.

We model the elementary link as a Werner state $\rho(w)$ with constant Werner parameter $w = w_0$ [23]:

$$\rho(w) = w |\Phi^+ \rangle \langle \Phi^+ | + (1 - w) \frac{1}{4}$$  \hspace{1cm} (1)

where the Bell state

$$|\Phi^+ \rangle = (|00\rangle + |11\rangle)/\sqrt{2}$$  \hspace{1cm} (2)

is a maximally-entangled two-qubit state and

$$\frac{1}{4} = (|00\rangle + |11\rangle) \otimes (|0\rangle + |1\rangle)/4$$

is the maximally-mixed state on two qubits. We refer to the parameter $w$ with $0 \leq w \leq 1$ as the Werner parameter. Since a Werner state is completely determined by its Werner parameter, we use the Werner parameter to indicate the quantum state.

Equivalently to the Werner parameter, we will also express the state’s quality using the fidelity, which for general density matrices $\rho$ and $\sigma$ is defined as

$$F(\rho, \sigma) := \text{Tr} \left( \sqrt{\sqrt{\rho} \sigma \sqrt{\rho}} \right)^2$$

The fidelity between a Werner state $\rho(w)$ and $|\Phi^+ \rangle \langle \Phi^+ |$ equals

$$F = \frac{1 + 3w}{4}$$

For the other 
PROTOCOL-UNIT:s, the success conditions are summarized in Table I. In short: we model entanglement swapping (SWAP) as succeeding with a constant probability $p_{\text{swap}}$. For entanglement distillation (DIST), we use the BBPSSW protocol [20] which we adapt by bringing the output state back into Werner form. The latter operation does not change the output state’s fidelity with the target state $|\Phi^+ \rangle$. The success probability $p_{\text{dist}}$ of distillation is a function of the input states’ Werner parameters (see [13] for details). The cut-off (CUT-OFF) success condition depends deterministically on the waiting time or the fidelity of the input links.

The states that any 
PROTOCOL-UNIT:s outputs are Werner states at any time of the execution of the protocol. Indeed, a successful entanglement swap or distillation attempt maps Werner states to Werner states (see [18] for a brief explanation). Also, the CUT-OFF leaves the input states untouched in case of success, thereby outputting Werner states if it got those as input. For each 
PROTOCOL-UNIT, the Werner parameters of the output links $w_{\text{out}}$ are a function of those of the input links and are given in Table I.

In addition to the fact that the 
PROTOCOL-UNIT:s change the quantum states they handle, the local quantum memories that are used to store the links are imperfect. In our model, a link

Fig. 2. Visualization of the waiting time until end-to-end entanglement is delivered for a 3-node repeater chain. The repeater scheme consists of the generation of two elementary links, followed by an entanglement swap on the two links. (a) A single link is generated in fixed-duration attempts, which succeed probabilistically and thus may fail (green line segment), after which generation is re-attempted until success (blue line segment). After that, the link is stored until it is consumed (dotted orange line segment). (b) A run of the 3-node protocol until the first swap attempt, which consists of first preparing two input links in parallel, followed by a Bell state measurement (BSM). The link that is generated earlier than the other needs to wait in the memory (link 1 in the figure, the ‘waiting’ is indicated by the dotted orange line). While waiting, the earlier link’s quality decreases due to decoherence. The total waiting time before the BSM equals the maximum of the generation times of the two links. The BSM operation can fail, in which case the two links are lost and need to be regenerated. (c) A full run of the 3-node protocol, consisting of failed entanglement swaps (green dashed box) on fresh links until the first successful swap (blue dashed box). The total waiting time is the sum of the waiting times for the parallel generation of each pair of elementary links, up to and including the first successful swap.
with initial Werner parameter $w$, which lives in memory for time $\Delta t$ until it is retrieved, decoheres to Werner parameter

$$w_{\text{decayed}} = w \cdot e^{-\Delta t/t_{\text{coh}}}.$$  

(3)

where $t_{\text{coh}}$ is the joint coherence time of the two involved memories.

For simplicity, we ignore the time needed for classical communication between the nodes in this work as well as the time to perform the local operations. The algorithm we provide can be easily extended to include these features, following the extension described in [18].

In summary, for a given composite protocol (including the cut-off condition $\tau$ or $w_{\text{cut}}$ for each CUT-OFF block), the simulation of the entanglement distribution process is determined by 4 hardware parameters: the success probability of elementary link generation $p_{\text{gen}}$, the swap success probability $p_{\text{swap}}$, the Werner parameter of the elementary link $w_0$ and the memory coherence time $t_{\text{coh}}$.

C. Waiting time and produced end-to-end state in repeater schemes using probabilistic components

In this work, we study the time until the first entangled pair of qubits is generated between the end nodes of the repeater chain (called ‘waiting time’ from here on) and the state’s quality, expressed as its Werner parameter (recall that the end-to-end state is a Werner state, see section [II-B]). Because the repeater chain protocols we study in this work are composed of probabilistic components, both the waiting time and the end-to-end state’s Werner parameter are random variables. For an illustration of the random behavior of the waiting time, see Fig. 2. We characterize the quality by the averaged Werner parameter of all states generated at the same time step $t$. The algorithm we present in this work computes the probability distribution $Pr(T = t)$ of the waiting time $T$ and the average Werner parameter $W(t)$ of the end-to-end state which is delivered at time $t$.

We finish this section by noting that by considering the average Werner parameter, we ignore the ‘history’ of a link, resulting in a suboptimal estimation of the fidelity of the states. To see this, consider for example the three-node protocol of Fig. 1(b). In this protocol, the following two series of events lead to an output entangled pair between nodes $A$ and $B$ at time $t = 10$: (i) all GEN blocks fail at each timestep $t < 10$ but succeed at time $t = 10$, after which all other PROTOCOL-UNITS also succeed immediately, (ii) the PROTOCOL-UNITS between $A$ and $M$ all succeed at time $t = 1$, while the GEN blocks between $M$ and $B$ succeed at time $t = 10$, followed by all other remaining PROTOCOL-UNITS also succeeding at time $t = 10$. In case (i), no entanglement has waited in memory, whereas in case (ii), the produced link between $A$ and $M$ has waited 10 timesteps and decohered in that time. By keeping track of the timestamps at which the several PROTOCOL-UNITS succeeded, one could distinguish these two scenarios. Since the resulting fidelity estimation computation is rather complex and in this work, we focus on quantifying the effect of a cut-off, we leave such advanced fidelity estimation for future work.

III. COMPUTING THE WAITING TIME DISTRIBUTION AND THE OUTPUT WERNER PARAMETER

In this section, we present closed-form expressions of the waiting time probability distribution and Werner parameter of the output links for each PROTOCOL-UNIT, as function of waiting time distribution and Werner parameter of its input links. Expressions for a composite protocol are obtained by iterative application over the PROTOCOL-UNITS that the protocol consists of. These expressions naturally lead to an algorithm for their evaluation, which we also present in this section.

Closed-form expressions for GEN and SWAP were already obtained by Brand et al. [18], who explicitly mentioned that their approach does not generalize straightforwardly to DIST. Here, we include DIST and even CUT-OFF, provided the latter is succeeded by SWAP or DIST. The novel idea is to use separate expressions for the waiting time probability distribution of a successful and failed attempt. We then express the total waiting time distribution and the Werner parameter as those of the successful attempt averaged by the occurrence probability of all possible sequences of failed attempts, where the weighted average is efficiently computed using convolution. As an additional benefit, the evaluation algorithm for SWAP is faster than the one presented by Brand et al.

In the following, we first derive general closed-form expressions for the waiting time distribution and Werner parameter of one PROTOCOL-UNIT in section [III-A]. We then give specific expressions for each PROTOCOL-UNIT individually in sections [III-B] to [III-E]. In the last section (section [III-F]), we show how these expressions can be converted into an efficient algorithm. We also explain how to modify the closed-form expressions using the discrete Fourier transform, motivated by its use in [24], [25]. These modified expressions lead to an even faster algorithm for computing the waiting time and Werner parameter, which we provide in Appendix [B]. We denote the random variables of the waiting time and average Werner parameter as $T$ and $W(t)$, with subscript $A$ and $B$ for the input links and ‘out’ for the output link (see Fig. 3).

A. General closed-form expressions for waiting time and produced states for all protocol-units

1) Random variable expression for the waiting time of PROTOCOL-UNITS: We start by presenting an expression for the random variable $T_{\text{out}}$. To study the waiting time distribution, we divide the total waiting time into the waiting time for each attempt. An attempt can fail or succeed and it repeats...
The expression in round brackets evaluates to 1 precisely if $Y^{(i)} = 1$ and $Y^{(j)} = 0$ for all $j < k$, and to 0 in all other cases. This factor thus makes that only the sum $\sum_{i=1}^{k} M^{(i)}$ is taken for which $k$ is the first successful attempt. Notice that $Y^{(j)}$ and $M^{(i)}$ are correlated for all $i = j$ because they describe the same attempts. In the next section, we go further to compute the probability distribution of $T_{\text{out}}$.

2) A closed-form expression for the waiting time distribution: In the following, we give an expression of the waiting time distribution $\Pr(T_{\text{out}} = t)$ for one PROTOCOL-UNIT.

We consider the generation time of a successful or failed attempt separately and use the joint distribution of $M$ and $Y$. We define the joint distribution that one attempt succeeds/fails and takes time $t$ as

$$P_{t}(t) := \Pr(M = t, Y = 1) = \sum_{t_{A},t_{B}:\max(t_{A},t_{B}) = t} \Pr(T_{A} = t_{A}, T_{B} = t_{B}) \cdot p(t_{A}, t_{B}),$$

and

$$P_{t}(t) := \Pr(M = t, Y = 0) = \sum_{t_{A},t_{B}:\max(t_{A},t_{B}) = t} \Pr(T_{A} = t_{A}, T_{B} = t_{B}) \cdot [1 - p](t_{A}, t_{B}).$$

In the above equation, we iterate over all possible combinations of the input links’ generation time $t_{A}, t_{B}$ that leads to a waiting time $t$ for this attempt.

With the definition (6) and (7), the sum of the waiting time for all attempts can be obtained by

$$\Pr(T_{\text{out}} = t) = \sum_{k=1}^{\infty} \left( \sum_{j=1}^{k-1} P_{t}^{(j)} \right) \ast P_{t}(t)$$

where $\ast$ is the notation for convolution and the sum over $k$ considers all the possible numbers of attempts. The notation $\sum_{j=1}^{k-1} P_{t}^{(j)}$ represents the convolution of $k - 1$ independent func-
tions \( P_i \). In the above equation, the discrete linear convolution is defined by

\[
[f_1 * f_2](t) = \sum_{t'=0}^{t} f_1(t - t') \cdot f_2(t').
\] (9)

If \( f_1, f_2 \) describe two probability distributions of two random variables, their convolution is the distribution of the sum of those two random variables. However, neither \( P_1 \) or \( P_2 \) characterizes a random variable since they are joint distributions including \( Y \). That is to say, \( P_2 \) and \( P_3 \) do not sum up to 1. Instead, we have

\[
\sum_t P_2(t) + \sum_t P_3(t) = 1.
\]

Therefore, the convolution here cannot be simply interpreted as a sum of two random variables. Instead, it is the summed waiting time conditioned on the success/failure of each attempt.

As we will show in sec. III-F eq. (8) is sufficient for the derivation of the main algorithm for computing the probability distribution of \( T_{out} \) we present in this work. The algorithm’s runtime is partially determined by the sum and the convolution in the summand in eq. (8). Fortunately, these can be eliminated by the use of the discrete Fourier transform, resulting in a faster alternative algorithm. Below, we use the Fourier transform to derive an equivalent expression to eq. (8). The alternative algorithm is given in Appendix B.

Since the discrete Fourier transform acts on a finite sequence of numbers, we first truncate the probability distribution at a fixed time \( L \), i.e. we obtain the finite sequence \( \{ \Pr(\tilde{T}_{out} = t)|t = 0, 1, 2, \ldots, L \} \). If \( \tilde{x} := x_0, x_1, \ldots, x_{L-1} \) is a sequence of complex numbers, then its Fourier transform \( \mathcal{F}(\tilde{x}) \) is the sequence \( y_0, y_1, \ldots, y_{L-1} \) given by

\[
y_j = \sum_{k=0}^{L-1} x_k \cdot \exp(-2\pi i \cdot j \cdot k/L)
\] (10)

where i is the complex unit. The Fourier transform is a linear map and moreover it converts convolutions into element-wise multiplication, i.e. \( \mathcal{F}(\tilde{x} \cdot \tilde{y}) = \mathcal{F}(\tilde{x}) \cdot \mathcal{F}(\tilde{y}) \). As a consequence, taking the Fourier transform of both sides of eq. (8) yields

\[
\mathcal{F}[\Pr(\tilde{T}_{out} = t)] = \sum_{k=1}^{\infty} \left[ \prod_{j=1}^{k-1} \mathcal{F}(P_{i})^{(j)} \right] \cdot \mathcal{F}[P_s](t).
\]

Because, \( P_{i}^{(j)} \) are identical distribution for all \( j \), we use the identity \( \sum_{k=1}^{\infty} x^{(k-1)} = 1/(1 - x) \) to obtain

\[
\Pr(\tilde{T}_{out} = t) = \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P_s]}{1 - \mathcal{F}[P_i]} \right](t).
\] (11)

3) A closed-form expression for the Werner parameter:

Here, we derive the expression for the Werner parameter \( W_{out}(t) \).

To arrive at \( W_{out}(t) \), we first compute the average Werner parameter of the output link of one attempt, given that it succeeds and finishes at time \( t \):

\[
W_s(t) = \frac{\sum \Pr(T_A = t_A, T_B = t_B) \cdot [p \cdot W_{out}](t_A, t_B)}{P_s(t)}.
\] (12)

Here, \( w_{out} \) is the Werner parameter of the output link of a successful attempt and \( p \) the success probability (Table I). We again simplify the notation with \( w_{out}(t_A, t_B) = w_{out}(t_A, t_B, W_{A}(t), W_{B}(t)) \).

Next, we take a weighted average of \( W_s^t \) over all possible sequences of failed attempts, followed by a single successful attempt:

\[
W_{out}(t) = \sum_{k=1}^{\infty} \left[ \prod_{j=1}^{k-1} P_{i}^{(j)} \right] \cdot (P_s \cdot W_{out})(t) \cdot \Pr(T_{out} = t).
\] (13)

where \( \bigstar \) \( P_{i}^{(j)} \) computes the waiting time distribution of \( k-1 \) failed attempts and the additional convolution is the weighted average.

For eq. (8), which is an expression for the probability distribution of \( T_{out} \), we obtained a more compact equivalent, eq. (11). By moving to Fourier space. By an analogous derivation, we can get a more compact expression for \( W_{out} \) than eq. (13):

\[
W_{out}(t) = \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P_s \cdot W_{out}]}{1 - \mathcal{F}[P_i]} \right](t).
\] (14)

B. Specific case: GEN

We give here the expression for PROTOCOL-UNIT GEN. Since GEN does not have input links, the output does not rely on the expression introduced in the section III-A. Because one attempt in GEN takes one time step and the success probability \( p_{gen} \) is a constant, the waiting time can be described by a geometric distribution

\[
\Pr(T_{out} = t) = p_{gen}(1 - p_{gen})^{t-1}.
\]

The output state is a Werner state with Werner parameter \( w_0 \) as described in section II-B.

C. Specific case: SWAP

For entanglement swap, since \( p_{swap} \) is constant, \( Y \) is not correlated with \( M \). As a result, \( P_s \) and \( P_t \) differ only by a constant coefficient (see 6 and 7). Therefore, we can factor the constant out and get

\[
\Pr(T_{out} = t) = \sum_{k=1}^{\infty} p_{swap}(1 - p_{swap})^{k-1} \left[ \prod_{j=1}^{k} m \right]
\]

where

\[
m(t) := \Pr(M = t) = \sum_{t_A, t_B = \max(t_A, t_B) = t} \Pr(T_A = t_A, T_B = t_B).
\]
By substituting

Compared to the expression in [18], this expression replaces

we can compute

obtain

This is exactly the geometric compound distribution obtained in [18].

For the Werner parameter, we can directly use (13) and obtain

\[ W_{out} = \sum_{k=1}^{\infty} p_{\text{swap}}(1 - p_{\text{swap}})^{k-1} \left[ \left( \begin{array}{l} k-1 \\ j=1 \end{array} \right) \ast (m \cdot W'_s) \right]. \]

(15)

Compared to the expression in [18], this expression replaces the iteration over all pairs of possible input Werner parameters for each \( k \) by convolution.

Both expressions above can also be written in Fourier space by substituting \( P_s = p_{\text{swap}}m(t) \) and \( P_1 = (1 - p_{\text{swap}})m(t) \) in (11) and (14).

D. Specific case: DIST

For entanglement distillation, the success probability depends on the Werner parameters. As discussed in section III-A, we can compute \( T_{out} \) and \( W_{out} \) because we iterate over all possible combinations of \( t_A \) and \( t_B \) and we use \( W(t) \) to reduce the dependence on Werner parameters to the dependence on the waiting time. The calculation goes as follows. First, we compute \( P_t \) and \( P_s \) using \( p(t_A, t_B) = p_{\text{dist}}(W(t_A), W(t_B)) \) (Table I). Then, we plug in \( P_t \) and \( P_s \) in (8) to compute \( T_{out} \). Finally, \( W_{out} \) can be calculated similarly using Table I (12) and (13).

E. Specific case: CUT-OFF

CUT-OFF selects the input links and accepts them if the cut-off condition described in section III-B is fulfilled. We consider only the case where CUT-OFF is followed by SWAP or DIST, so that the two blocks together output a single entangled link.

1) The waiting time distribution: We define a new binary variable \( Y_{cut} \) representing whether the cut-off condition is fulfilled. The corresponding success probability is described by \( p_{cut} \) in Table I. In addition, we also define the waiting time of one cut-off attempt as \( Z \), in contrast to \( M \) for a swap or distillation attempt. For CUT-OFF, we need to distinguish the waiting time of a successful and a failed attempt. In the case of success, we always have \( Z_s = \max(T_A, T_B) \), i.e. we wait until two links are produced. However, in the case of failure, the waiting time is different for different cut-off strategies. With the notation \( Z_{f} = t_{\text{fail}}(T_A, T_B) \), we have the following: for DIF-TIME-CUT-OFF, \( t_{\text{fail}}(T_A, T_B) = \min(T_A, T_B) + \tau \), because there is no need to wait for the second link longer than the cut-off threshold. For MAX-TIME-CUT-OFF, \( t_{\text{fail}}(T_A, T_B) \) is the constant \( \tau \), i.e. the maximal allowed waiting time. For FIDELITY-CUT-OFF, it is \( t_{\text{fail}}(T_A, T_B) = \max(T_A, T_B) \).

Similar as the nested structure shown in Fig. 3 a swap or distillation attempt is now composed of several cut-off attempts. We can write its waiting time \( M \) as

\[ M = \sum_{k} \left\{ \left[ \prod_{j=1}^{k-1} (1 - Y_{cut}^{(j)}) \right] \ast \left[ Z_s^{(k)} + \sum_{i=1}^{k-1} (Z_{f}^{(i)}) \right] \right\} \]

This expression will replace \( M = \max(T_A, T_B) \) used in (5).

For \( \tau = \infty \) or \( w_{cut} = 0 \), i.e. no cut-off, \( Y_{cut} \) is always 1.

Therefore, \( k = 1 \) is the only surviving term and the two expressions coincide.

To calculate the waiting time distribution, we need three joint distributions: \( P'_f \) for unsuccessful input link preparation because of the cut-off, \( P'_{\text{cuts}} \) for successful preparation but unsuccessful swap/distillation and \( P'_{s,s} \) for both successful:

\[ P'_f(t) = \Pr(M = t, Y_{cut} = 0) = \sum_{t_A, t_B} \Pr(T_A = t_A, T_B = t_B) \cdot [1 - p_{cut}(T_A, T_B)] \]

\[ P'_{s,s}(t) = \Pr(M = t, Y_{cut} = 1, Y = 0) = \sum_{t_A, t_B} \Pr(T_A = t_A, T_B = t_B) \cdot [p_{cut} \cdot (1 - p)](t_A, t_B). \]

\[ P'_{s,s}(t) = \Pr(M = t, Y_{cut} = 1, Y = 1) = \sum_{t_A, t_B} \max_{\{t_A, t_B\} = t} \Pr(T_A = t_A, T_B = t_B) \cdot [p_{cut} \cdot p](t_A, t_B). \]

The prime notation indicates that they describe the waiting time of one attempt in CUT-OFF, in contrast to one attempt in SWAP or distillation.

For one attempt in SWAP/distillation with time-out, we then get similarly to (8)

\[ P_s(t) = \Pr(M = t, Y = 1) = \sum_{k} \left[ \left( \begin{array}{c} k-1 \\ j=1 \end{array} \right) \ast P_f^{(j)} \right] (t) \]

\[ P_t(t) = \Pr(M = t, Y = 0) = \sum_{k} \left[ \left( \begin{array}{c} k-1 \\ j=1 \end{array} \right) \ast P'_s \right] (t) \]

as well as the expressions in Fourier space analogous to (11)

\[ P_s(t) = \Pr(M = t, Y = 1) = \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P'_{s,s}]}{1 - \mathcal{F}[P'_f]} \right]. \]

\[ P_t(t) = \Pr(M = t, Y = 0) = \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P'_{s,f}]}{1 - \mathcal{F}[P'_f]} \right]. \]

The total waiting time then follows by substituting the expressions for \( P_t \) and \( P_s \) above in (8) or (11).

For entanglement swap, i.e. constant success probability \( p_{\text{swap}} \), simplification can be made for this calculation. In this special case, \( P'_{s,f} \) and \( P'_{s,s} \) differ only by a constant and the same holds for \( P'_s \) and \( P'_t \).

2) The Werner parameter: For the Werner parameter, we now need three steps.

Next, we compute the Werner parameter \( W_s(t) \) as a function of time \( t \) that includes the failed cut-off attempts, in analogy to the derivation of (13). \( W_s(t) \) is the Werner parameter that pair of output links of CUT-OFF will produce, given that the swap or distillation operation following is successful:

\[ W_s(t) = \frac{ \sum_{k=1}^{\infty} \left[ \left( \begin{array}{c} k-1 \\ j=1 \end{array} \right) \ast (P'_f \ast \mathcal{F}[P'_{s,s} \cdot W'_s]) \right] }{ P'_f(t) }. \]
Finally, we consider the time consumed by failed attempts in SWAP or DIST and obtain

\[ W_{\text{out}}(t) = \sum_{k=1}^{\infty} \left[ \frac{1}{k^{j=1}} P_j \ast (P_s \cdot W_s) \right](t) \cdot \Pr(T_{\text{out}} = t). \]

Using the Fourier transform, the two expressions above become

\[ W_s(t) = \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P_{s}', W_{s}']}{1 - \mathcal{F}[P_s]} \right] \cdot \mathcal{F}^{-1} \left[ \frac{\mathcal{F}[P_s \cdot W_s]}{1 - \mathcal{F}[P_s]} \right] \cdot \Pr(T_{\text{out}} = t). \]

**F. Converting the closed-form expressions into an efficient algorithm**

In the sections above, we presented closed-form expressions for \( T_{\text{out}} \) and \( W_{\text{out}} \) for each of the four PROTOCOL-UNITs, as a function of waiting time distribution and Werner parameter of the input links. In order to convert these expressions into an algorithm, we take the same approach as in \[18\] and cap the infinite sum in \((8)\) and \((13)\) by a pre-specified truncation time \( t_{\text{trunc}} \). This yields a correct \( \Pr(T_{\text{out}} = t) \) and \( W_{\text{out}}(t) \) for \( t \in \{1, \ldots, t_{\text{trunc}}\} \) since in each of the expressions with an infinite sum above, \( \Pr(T_{\text{out}} = t) \) and \( W_{\text{out}}(t) \) are only dependent on waiting time and Werner parameter of input links produced at time \( t' \leq t \).

We now show that the algorithm scales polynomially in terms of \( t_{\text{trunc}} \). To analyze the complexity, we divide the algorithm into two parts: computing the distribution for one attempt, \( t_{\text{trunc}} \), and for the whole PROTOCOL-UNIT (\( t_{\text{trunc}} \) and \( t_{\text{trunc}} \)). The complexity for the first part is \( O(t_{\text{trunc}}^2) \) since it iterates over two discrete random variables up to \( t_{\text{trunc}} \). For the second part, because we need at least one time step in each attempt, i.e. \( \Pr(T = 0) = 0 \), only the first \( t_{\text{trunc}} \) convolutions will have non-zero contribution. We can perform the convolution iteratively for each \( k \) using at most \( t_{\text{trunc}} \) convolutions. The complexity of one convolution with fast Fourier transform (FFT) is \( O(t_{\text{trunc}} \log t_{\text{trunc}}) \) \[26\]. Thus, the complexity of the second part scales as \( O(t_{\text{trunc}} \log t_{\text{trunc}}) \). The overall complexity, therefore, is \( O(t_{\text{trunc}}^2 \log t_{\text{trunc}}) \).

In appendix \[B\] we show that with further simplification of \((6)\) and \((7)\) as well as expressions in Fourier space (equations \((11)\) and \((13)\)), the complexity can be reduced to \( O(t_{\text{trunc}} \log t_{\text{trunc}}) \), with an exponentially vanishing error.

The preceding discussion shows that the algorithm is efficient as a function of the truncation time. However, for fixed truncation time, the probability mass captured by the algorithm decreases as the number of nodes increases. For protocols without cut-off, variations of the arguments in \[18\] would allow to prove that the algorithm introduced here is also efficient for fixed probability mass. Unfortunately, the arguments do not translate to protocols with cut-off. This is because for these protocols, the truncation time that covers a fixed probability mass can grow exponentially with the number of nodes, i.e. such an algorithm can not exist.

As an example, consider a nested protocol on \( 2^n \) repeater segments \((n = 0, 1, 2, \ldots)\), which for \( n = 1 \) consists of a GEN block only, and for each additional level \( n > 1 \), each pair of adjacent links is connected by a CUT-OFF followed by a SWAP. We set \( \tau = 0 \) for each cut-off, i.e. all elementary links need to be generated at the same time and all entanglement swaps should succeed at the first attempt for the links to survive all the cut-offs. Since \( 2^n \) elementary links need to be generated and the protocol consists of \( 2^n - 1 \) swaps, the probability of successful end-to-end entanglement before time \( t \) equals \( 1 - (1 - p)^t \) with \( p = p_{\text{gen}}^{n-1} p_{\text{swap}}^{n-2} \), i.e. decreases exponentially in the number of nodes \( N = 2^n + 1 \).

**IV. Optimization**

In this section, we describe the details of our optimization over cut-offs, including the figure of merit and optimization method.

In our numerical study, we use the secret-key rate of the BB84 protocol \[27\] as a figure of merit to assess the performance of composite repeater protocols. We compute the secret-key rate \( R \) as the secret-key fraction divided by the average waiting time

\[ R = \frac{r}{T}. \]

The secret-key fraction \( r \) describes the amount of secret key that can be extracted from the generated entanglement and is given by \[28, 29\]

\[ r(w) = \max \{0, 1 - h(e_X(w)) - h(e_Z(w))\} \]

where \( h(p) = -p \log_2(p) - (1 - p) \log_2(1 - p) \) is the binary entropy function and \( e_X \) (\( e_Z \)) is the quantum bit error rate in the X (Z) basis. Since the quantum states tracked by our algorithm are Werner states at any point in the execution of the composite repeater protocol (see section \[II\]), the quantum bit error rate can be expressed as function of the end-to-end state’s Werner parameter:

\[ e_Z(w) = \langle 01 | \rho(w) | 01 \rangle + \langle 10 | \rho(w) | 10 \rangle = 1 - \frac{w}{2} \]

for a Werner state \( \rho(w) \) defined in \[4\]. The same result holds for \( e_X \) because of the symmetry of the Werner state. In section \[C\] we detail how we compute the secret-key rate with truncated waiting time distribution and Werner parameter obtained from the algorithm in section \[III-F\].

Since we have discrete time steps, we need an optimization algorithm which is compatible with a discrete search space. We choose the differential evolution algorithm implemented in the SciPy-optimization library of the Python programming language \[30, 31\].

**V. Numerical results**

In this section, we optimize over repeater protocols with cut-offs in order to maximize the rate at which secret key can be extracted from the produced end-to-end entanglement. First, we use our algorithm from section \[III\] and the DIF-TIME-CUT-OFF strategy (section \[II\]) to study the effect of the cut-off on the waiting time and fidelity and show that the use of a cut-off boosts secret-key rate. We then extend
We validate correctness of the implementation by comparison with an extended version of the Monte Carlo algorithm (dots), which we use for validating the correctness of our implementation (see appendix A for details).

We start by investigating the effect of cut-off strategy, MAX-TIME-CUT-OFF and FIDELITY-CUT-OFF, and compare their performance. For all three cut-off strategies, we observe that the resulting repeater protocols produce secret key at significantly higher rates than their no-cut-off alternatives. Finally, we focus on the DIF-TIME-CUT-OFF strategy and analyze the sensitivity of the optimal cut-off threshold with respect to the hardware parameters.

We investigate repeater protocols with 3 nesting levels where at each nesting level the range of entanglement is doubled by an entanglement swap. The protocol thus spans $2^3 = 8$ segments ($8 + 1 = 9$ nodes). Each entanglement swap operation is preceded by a cut-off, i.e., the scheme is of the form

$$\text{GEN} \rightarrow (\rightarrow \text{CUT-OFF} \rightarrow \text{SWAP})^3.$$  \hfill (17)

The numerical results in this section were obtained using our open-source implementation \cite{2020arXiv200501255S} of the algorithm from section III on consumer-market hardware (Intel i7-8700 CPU). We validated correctness of the implementation by comparison with an extended version of the Monte Carlo algorithm from \cite{2020arXiv200501255S} (see Fig. 4 and appendix A for details).

A. Effect of DIF-TIME-CUT-OFF on the waiting time and fidelity

We start by investigating the DIF-TIME-CUT-OFF strategy, where links are discarded if their production times differ by more than a predetermined threshold $\tau$. We compute waiting time and average fidelity for a particular choice of the cut-off threshold at each of the three levels and compare it with the protocol without cut-off (cut-off duration $\tau = \infty$ at each nesting level), see Fig. 4. We observe that the cut-off increases fidelity at the cost of longer waiting time, as one would intuitively expect. We further quantify the time-fidelity trade-off for a range of cut-offs in Fig. 5. For maximizing the secret key rate, we observe a single optimal choice of the cut-off threshold $\tau$. 

B. Extension to other cut-off strategies

We extend the analysis of the previous sub-section to two other cut-off strategies: a cut-off on the fidelity (FIDELITY-CUT-OFF) and on the total waiting time (MAX-TIME-CUT-OFF, see section III and Table 1 for definitions). To be precise, we choose the same 9-node protocol from (17) and use FIDELITY-CUT-OFF and MAX-TIME-CUT-OFF as the cut-off unit, respectively.

We observe that a single optimal cut-off threshold exists for both strategies, as we saw before already for the DIF-TIME-CUT-OFF strategy in Fig. 5. For each strategy, we optimize their cut-off parameters and plot the waiting time distribution and fidelity distribution in Fig. 6. As shown in the figure, although the FIDELITY-CUT-OFF yields the highest secret-key rate, the distribution and resulting secret-key rate of the DIF-TIME-CUT-OFF strategy are very close to those of the FIDELITY-CUT-OFF strategy. In contrast, the MAX-TIME-CUT-OFF strategy performs significantly worse in the achieved secret-key rate ($\approx 10\%$). We find similar behavior also in other parameter regimes.

Since the DIF-TIME-CUT-OFF strategy is straightforward to implement in experiments while it performs only marginally worse than the best of the three strategies (FIDELITY-CUT-OFF), we focus on this strategy for further analysis.
C. Performance of the optimal cut-off for varying hardware parameters

We proceed with optimizing the cut-off in the DIF-TIME-CUT-OFF strategy to maximize the secret key rate for a range of parameters. The maximal secret-key rates for different repeater parameters are shown in Fig. 7(a-d). We observe that cut-offs extend the parameter regime for which secret key can be generated. To see how much one can gain in the secret key rate by using cut-offs, we choose two parameters $t_{coh}$ and $w_0$ and plot the absolute increase in Fig. 8. We observe that the use of the optimal cut-off increases the secret key rate for the entire parameter range plotted and the improvement is largest close to the threshold parameters at which the no-cut-off protocol starts to produce nonzero secret key.

In addition, we compare uniform and non-uniform cut-offs, where ‘uniform’ means that we choose the same cut-off time for each nesting level. For the parameter regimes studied, we observe that non-uniform and uniform cut-off perform similarly, see Fig. 7(a-d).

Our next step is the sensitivity analysis of cut-off performance in the hardware parameters. For this, we first choose baseline values for the four hardware parameters and find the corresponding optimal cut-off $\tau_{baseline}$. Given a target set of parameters that deviates slightly from the baseline values (optimal cut-off $\tau_{target}$), we quantify the sensitivity by their relative performance difference

$$R(\tau_{target}) - R(\tau_{baseline}) = \frac{R(\tau_{target})}{R(\tau_{target})}$$

where $R$ is the secret-key rate achieved by the repeater protocol. If this relative difference is small, the performance of cut-off is insensitive to the parameter deviation.

In Fig. 8(e-h), we plot the relative performance difference for deviations in each of the four hardware parameters separately. We find that the performance of the baseline cut-off is influenced most by variation in coherence time, while it is largely insensitive to change in the swap success probability. For the coherence time and the remaining two parameters, the elementary link quality and the success probability of elementary link generation, we distinguish the case where the parameter is improved and the regime where the parameter is made worse. We observe that a worse parameter results in a significant performance difference with the optimal cutoff, while the performance difference is small when the parameter is improved.

We finish by investigating the most influential parameter, the coherence time, in Fig. 9. We observe that the optimal threshold depends approximately linearly on the memory coherence time, which could serve as a heuristic for choosing a performant cut-off.

VI. CONCLUSION

In this work, we optimized the secret key rate over repeater protocols including cut-offs. Our main tool is an algorithm for computing the probability distribution of waiting time and fidelity of the first generated end-to-end link. The algorithm is applicable to a large class of quantum repeater schemes that can include cut-off strategies and distillation. Its runtime is polynomial in the support size of the probability distribution of waiting time.

Our simulations show that the use of the optimal cut-off lowers the hardware quality threshold at which secret key can be generated compared to the no-cut-off alternative. Furthermore, we observed an increase in secret-key rate for the entire regime studied for which the no-cut-off protocol produces nonzero key.

Regarding the choice of cut-off, we find that uniform cut-offs lead to a negligible reduction in the secret key rate compared to the optimal set of cut-offs which differ per nesting level. Moreover, the optimal uniform cut-off is highly sensitive to the quality of the memory, while it is barely influenced by
the success probability of swapping. Such sensitivity could guide the heuristic cut-off optimization of more complex protocols.

**APPENDIX A**

**Validation Against a Monte Carlo Algorithm**

In this section, we verify that our implementation of the deterministic algorithm presented in section III is correct by validation against the Monte Carlo sampling algorithm from Brand et al. [18]. For all repeater schemes we ran (up to $2^{10} + 1$ nodes for some parameters), we observed good agreement between the waiting time probability distribution and Werner parameter the algorithms computed, which is convincing evidence that our implementation is correct. Fig. 7 depicts the result of a typical run.

What follows is a brief description of the Monte Carlo algorithm from Brand et al. [18], including an extension to CUT-OFF. Each run of the Monte Carlo algorithm samples a tuple of waiting time and Werner parameter. It is defined recursively by having a dedicated function for each PROTOCOL-UNIT (described below) call the dedicated functions of the two PROTOCOL-UNITS that produce its two input links. The recursion follows the repeater protocol’s tree structure (see Fig. 1), resulting in a sampling algorithm of waiting time and Werner parameter of the entire repeater protocol.

The dedicated functions for each of the four PROTOCOL-UNITS are as follows. If the protocol is only a GEN, the Monte Carlo algorithm samples the waiting time from the geometric distribution with parameter $p_{gen}$ and the Werner parameter is the constant $w_0$. For the other PROTOCOL-UNITS, each of which takes two links as input, the algorithm begins by initializing the total elapsed time $t = 0$. Then, it enters a loop which starts by calling the dedicated functions of the PROTOCOL-UNITS that produce the two input links, resulting in two samples $(t_A, w_A)$ and $(t_B, w_B)$. The algorithm randomly declares ‘success’ or ‘failure’ according to the success probability in Table I. If it succeeds, the function breaks the loop and outputs $t + \max(t_A, t_B)$ and the resulting Werner parameter $w_{out}(t_A, w_A, t_B, w_B)$ (see Table I). If it fails, the total elapsed time $t$ is increased by the waiting time $(\max(t_A, t_B)$ for SWAP and DIST, $\min(t_A, t_B) + \tau$ for CUT-OFF$)$ and the function goes back to the start of the loop.

**APPENDIX B**

**Alternative Algorithm and Its Complexity**

In section III-F we presented an $O(t_{trunc}^2 \log t_{trunc})$ algorithm for evaluating analytically-derived expressions for the waiting time distribution and average fidelity. Here, we outline how the algorithm can be modified to achieve a complexity reduction to $O(t_{trunc} \log t_{trunc})$ for protocols composed
of PROTOCOL-UNITS in Table I except for FIDELITY-CUT-OFF. Similar to the algorithm from the main text, the modified algorithm consists of two steps: first, evaluating the expressions regarding a single attempt (equations (6), (7) and (12)), followed by computing expressions regarding the whole PROTOCOL-UNIT (equations (11) and (14)). We show a complexity reduction for both.

For the first part, we show how to evaluate (6), (7) and (12) in time $O(t_{\text{trunc}})$, improving on the $O(t_{\text{trunc}}^2)$ runtime of the algorithm in the main text. Our insight here is that $p$ and $p^* w_{\text{out}}$, for SWAP and DIST (see Table I), can always be written in the form

$$\sum_i f^{(i)}(t_A) \cdot g^{(i)}(t_B)$$

(19)

where the $f^{(i)}$ and $g^{(i)}$ are arbitrary functions on the real numbers. For instance, given $t_A \geq t_B$, we can write the success probability of distillation $p_{\text{dist}}$ with $f^{(1)}(t_A) = \frac{1}{2}$, $g^{(1)}(t_B) = 1$ and $f^{(2)}(t_A) = \frac{1}{2} p_{\text{swap}} w_A(t_A) \exp(-\frac{t_A}{t_{\text{coh}}})$, $g^{(2)}(t_B) = w_B(t_B) \exp(\frac{t_B}{t_{\text{coh}}})$. Consequently, each of (6), (7) and (12) can be written in the form

$$\sum_{t_A, t_B; \text{max}(t_A, t_B) = t} \sum_i f^{(i)}(t_A) g^{(i)}(t_B)$$

(20)

which can be rewritten by splitting up the sum in the regime $t_A \geq t_B$ and $t_B > t_A$:

$$\sum_{t_B = 0}^{t_B = t} \sum_{t_A = 0}^{t_A = t-1} \sum_i f^{(i)}(t) g^{(i)}(t_B) + \sum_{t_A = t_B; \text{max}(t_A, t_B) = t} \sum_i f^{(i)}(t_A) g^{(i)}(t).$$

(21)

The first term in (21) can be written as

$$\sum_{t_B = 0}^{t_B = t} \sum_{t_A = 0}^{t_A = t-1} f^{(i)}(t) \cdot G^{(i)}(t)$$

(22)

where we have defined

$$G^{(i)}(t) = \sum_{t_B = 0}^{t_B = t} \Pr(T_B = t_B) g^{(i)}(t_B).$$

The expression for the second term in (21) can be found analogously. Computing (22) for all $t$ is now performed by first computing $G^{(i)}(t)$ for all $t$, which requires linear time in $t_{\text{trunc}}$, and then evaluating (22) for fixed $t$ in constant time. Therefore, the complexity for computing (22) and also for (20) for all $t$ scales as $O(t_{\text{trunc}})$.

This complexity holds also for protocols with DIF-TIME-CUT-OFF and MAX-TIME-CUT-OFF, as the cut-off condition appears only as an additional constraint on $t_A$ and $t_B$ in the sum of (21). For the third cut-off strategy we consider in this work, FIDELITY-CUT-OFF, the cut-off condition is not a function of time and therefore the above method does not work.

The second part regards the evaluation of (8) and (13) which is done exactly by the algorithm from the main text in time $O(t_{\text{trunc}}^2 \log t_{\text{trunc}})$. Here, we give an $O(t_{\text{trunc}} \log t_{\text{trunc}})$-algorithm which evaluates the equivalent expressions in Fourier space given in section III-C (equations (11) and (14)) with arbitrarily small error. We proceed in two steps. First, we show how to evaluate the expressions in Fourier space exactly in time $O(t_{\text{trunc}} \log t_{\text{trunc}}^2)$. Then, we show how to achieve a reduction to $O(t_{\text{trunc}} \log t_{\text{trunc}})$ with an arbitrarily small error.

The expressions in Fourier space (equations (11) and (14)) hold for any $t$ in case $P_s$, $P_f$ and $W_i$ are defined for all $t \geq 0$. For the first part, we show how to evaluate (6), (7) and (12) in time $O(t_{\text{trunc}})$, improving on the $O(t_{\text{trunc}}^2)$ runtime of the algorithm in the main text. Our insight here is that $p$ and $p^* w_{\text{out}}$, for SWAP and DIST (see Table I), can always be written in the form

$$\sum_i f^{(i)}(t_A) \cdot g^{(i)}(t_B)$$

(19)

where the $f^{(i)}$ and $g^{(i)}$ are arbitrary functions on the real numbers. For instance, given $t_A \geq t_B$, we can write the success probability of distillation $p_{\text{dist}}$ with $f^{(1)}(t_A) = \frac{1}{2}$, $g^{(1)}(t_B) = 1$ and $f^{(2)}(t_A) = \frac{1}{2} p_{\text{swap}} w_A(t_A) \exp(-\frac{t_A}{t_{\text{coh}}})$, $g^{(2)}(t_B) = w_B(t_B) \exp(\frac{t_B}{t_{\text{coh}}})$. Consequently, each of (6), (7) and (12) can be written in the form

$$\sum_{t_A, t_B; \text{max}(t_A, t_B) = t} \sum_i f^{(i)}(t_A) g^{(i)}(t_B)$$

(20)

which can be rewritten by splitting up the sum in the regime $t_A \geq t_B$ and $t_B > t_A$:

$$\sum_{t_B = 0}^{t_B = t} \sum_{t_A = 0}^{t_A = t-1} \sum_i f^{(i)}(t) g^{(i)}(t_B) + \sum_{t_A = t_B; \text{max}(t_A, t_B) = t} \sum_i f^{(i)}(t_A) g^{(i)}(t).$$

(21)

The first term in (21) can be written as

$$\sum_{t_B = 0}^{t_B = t} \sum_{t_A = 0}^{t_A = t-1} f^{(i)}(t) \cdot G^{(i)}(t)$$

(22)

where we have defined

$$G^{(i)}(t) = \sum_{t_B = 0}^{t_B = t} \Pr(T_B = t_B) g^{(i)}(t_B).$$

The expression for the second term in (21) can be found analogously. Computing (22) for all $t$ is now performed by first computing $G^{(i)}(t)$ for all $t$, which requires linear time in $t_{\text{trunc}}$, and then evaluating (22) for fixed $t$ in constant time. Therefore, the complexity for computing (22) and also for (20) for all $t$ scales as $O(t_{\text{trunc}})$.

This complexity holds also for protocols with DIF-TIME-CUT-OFF and MAX-TIME-CUT-OFF, as the cut-off condition appears only as an additional constraint on $t_A$ and $t_B$ in the sum of (21). For the third cut-off strategy we consider in this work, FIDELITY-CUT-OFF, the cut-off condition is not a function of time and therefore the above method does not work.

The second part regards the evaluation of (8) and (13) which is done exactly by the algorithm from the main text in time $O(t_{\text{trunc}}^2 \log t_{\text{trunc}})$. Here, we give an $O(t_{\text{trunc}} \log t_{\text{trunc}})$-algorithm which evaluates the equivalent expressions in Fourier space given in section III-C (equations (11) and (14)) with arbitrarily small error. We proceed in two steps. First, we show how to evaluate the expressions in Fourier space exactly in time $O(t_{\text{trunc}} \log t_{\text{trunc}}^2)$. Then, we show how to achieve a reduction to $O(t_{\text{trunc}} \log t_{\text{trunc}})$ with an arbitrarily small error.

The expressions in Fourier space (equations (11) and (14)) hold for any $t$ in case $P_s$, $P_f$ and $W_i$ are defined for all $t \geq 0$.
However, in the implementation, we truncate the distribution and only have access to them for $0 \leq t < t_{\text{trunc}}$, each stored as an array of length $t_{\text{trunc}}$, and use the discrete Fourier transform defined in (10). The convolution defined in this way is a circular convolution:

$$[f_1 \ast f_2](t) = \sum_{t' = 0}^{t} f_1(t - t') \cdot f_2(t') + \sum_{t' = 1}^{L - 1} f_1(L + t - t') \cdot f_2(t'),$$

where $L$ is the length of the array and $\ast$ denotes the circular convolution.

The circular convolution introduces discrepancy compared to the linear convolution defined in (9) because $[f_1 \ast f_2](t) = [f_1 \ast f_2](t) + [f_1 \ast f_2](L + t)$. To avoid this, we pad the arrays of $P_s$, $P_f$ and $W_s$ with zeroes until a length of $L = t_{\text{trunc}}^2$, which is longer than the size of $t_{\text{trunc}}$ times convolution of arrays of size $t_{\text{trunc}}$ (see equivalent expressions (8) and (13), and the algorithm presented in section III-F). That is, we set $P_s(t) = 0$ and $P_f(t) = 0$ for $t_{\text{trunc}} \leq t < L = t_{\text{trunc}}^2$.

With this setup, the summand in the circular convolution is always 0 for $t' > t$ and it coincides with the linear one. The complexity of the obtained algorithm evaluating (11) and (14) is dominated by one Fourier transform and one inverse Fourier transform on an array of length $O(t_{\text{trunc}}^2)$. Since a Fourier transform on an array of length $L$ can be performed in time $O(L \log L)$, the algorithm has a complexity of $O(t_{\text{trunc}}^2 \log t_{\text{trunc}}^2)$.

We now show that we can reduce this complexity by zero-padding the arrays only until a length of $Ct_{\text{trunc}}$ for some predefined constant $C$, yielding an exponentially small error

$$\epsilon = \max_t \left( |\Pr(T_{\text{out}} = t) - \Pr(T_{\text{approx}} = t)| \right)$$

in $C$ of the distribution $\Pr(T_{\text{approx}} = t)$ obtained with circular convolution. The resulting algorithm has complexity of $O(Ct_{\text{trunc}} \log(Ct_{\text{trunc}})) = O(t_{\text{trunc}} \log t_{\text{trunc}})$. The motivation behind this reduction is that $\Pr(T_{\text{out}} = t)$ is the sum of all possible sequences of failed attempts (see (8)) and is exponentially decreasing for large $t$. For a fixed number of attempts $k$, the probability results from a successful attempt after at least $k - 1$ failed attempts. Therefore, it has an occurrence probability of at most $(1 - p)^{k-1}$, where $p$ is the success probability for a PROTOCOL-UNIT. To see this mathematically, we use the Young’s convolution inequality (33) and obtain

$$\left\| \sum_{j=1}^{k-1} P_t^{(j)} \ast P_s \right\| \leq \left\| P_t \right\|^{k-1} \left\| P_s \right\| \leq (1 - p)^{k-1}$$

where the norm is defined by $\|f(t)\| = \sum_t f(t)$. In addition, note that

$$\left\| \sum_{j=1}^{k-1} P_t^{(j)} \ast P_s \right\| (t) = 0 \quad \text{for} \quad t \geq kt_{\text{trunc}}$$

because $P_t(t)$ and $P_s(t)$ are finite arrays of length $t_{\text{trunc}}$. Hence, for $t \geq Kt_{\text{trunc}}$, we only need to consider the terms with $k \geq K+1$, i.e. cases with at least $K$ failed attempts. As a result, we obtain a bound for the probability given in (8) for $t \geq Kt_{\text{trunc}}$:

$$\Pr(T_{\text{out}} = t) \leq \sum_{k=K+1}^{\infty} (1 - p)^{k-1} = \frac{(1 - p)^K}{p}.$$
Fig. 10. Computation time of the algorithm from appendix B as a function of the number of nodes in the repeater chain using consumer-market hardware (Intel i7-8700 CPU). We plot the computation time for three different $\psi_{swap}$ and for protocols of the form GEN $\rightarrow$ (CUT-OFF $\rightarrow$ SWAP)$^n$, similar to [17], where $n$ is the nesting level and the number of nodes is $2^n + 1$. The truncation time is chosen such, that 99% of the probability mass is covered. Note that the plot's axes are both given in logarithmic scale; in such a log-log plot, a polynomial function is represented as a line. The used cut-off strategy is DIFF-TIME-CUT-OFF and the other parameters used are: $\psi_{gen} = 0.1$, $w_{coh} = 1.0$, $\psi_{swap} = 500/\psi_{swap}$, $\tau = 42/\psi_{swap}$. In this plot, the number of truncation time steps goes up to about $10^6$.

Accordingly, the average Werner parameter is an average over the successful execution

$$W = \sum_{t=1}^{t_{trunc}} \frac{W(t) \cdot Pr(T = t)}{Pr(T \leq t_{trunc})}.$$  

With the above equations, we calculate the secret-key rate defined in (16). In this work, we choose heuristically a $t_{trunc}$ such that $Pr(T \leq t_{trunc}) \geq 99\%$. With this choice, the difference in the secret key rate between protocols with finite and infinite $t_{trunc}$ is negligibly small.

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