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Published in:
Physical Review A (Atomic, Molecular, and Optical Physics)

Document Version:
Peer reviewed version

Queen's University Belfast - Research Portal:
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Non-Markovianity and System-Environment Correlations in a Microscopic Collision Model

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(Dated: April 8, 2014)

We show that the use of a recently proposed iterative collision model with inter-environment swaps displays a signature of strongly non-Markovian dynamics that is highly dependent on the establishment of system-environment correlations. Two models are investigated; one in which such correlations are cancelled iteratively and one in which they are kept all across the dynamics. The degree of non-Markovianity, quantified using a measure based on the trace distance, is found to be much greater for all coupling strengths, when system-environment correlations are maintained.

One of the obstacles to the grounding of fully scalable quantum technologies is, notoriously, the subversion of decoherence, that is the process of losing the information encoded in a quantum mechanical system due to its interaction with an environment [1, 2]. The uncontrollable process of exchange of information between a quantum system and its environment is responsible for the consequent degradation of the (quantum) coherence in the state of the former. The study of quantum decoherence has been the focus of a considerable body of investigations, both at the theoretical and experimental level, aimed at understanding and ultimately taming the effects that system-environment interactions have on a given dynamics.

More recently, inspired also by the improved control over small-scale solid-state quantum devices, whose open-system dynamics is made difficult by the non-flat inherent structure of the environment that surrounds them, considerable attention has been given to the characterisation of the fundamental differences between Markovian and non-Markovian open system dynamics [2, 3], a task that has been empowered by the formulation of useful tools for the quantitative assessment of the features of non-Markovianity [4, 5] which have enabled related investigations in various directions [6], including test-bed simulations of explicitly non-Markovian evolutions and the experimental study on the transition from one dynamical regime to the other [7].

The idea behind most of the proposed measures of non-Markovianity is the quantification of a “backflow” of some type. If the environment is able to retain and feed-back to the system it is interacting with part (or all) of the “information” (intended in a very broad sense) that the latter has previously poured into the former during the evolution, quantitative figures of merit can be identified that are able to signal the resulting dynamics as non-Markovian [4, 5]. One such indicator [4] is based on the use of the trace distance $D[\rho_1, \rho_2]$ [8, 9], which quantifies the degree of distinguishability between two arbitrary quantum states $\rho_1$ and $\rho_2$. When subjected to the same physical open-system model, two different states would in general undergo different evolutions. Trace distance is contractive under positive trace preserving maps, which is a property that enters significantly into the definition of non-Markovian quantum dynamics put forward in Ref. [4]: any increase in the trace distance should be taken as a fingerprint of non-Markovianity.

Albeit a unified view on the reasons behind the emergence of non-Markovian features has not yet been found, progresses have been made in the establishment of a hierarchical relation among some of the measures proposed so far [10, 11] and a careful assessment of the trace distance-based measure has allowed to pinpoint the emergence of system-environment correlations (SECs) established across the dynamics as being key for non-Markovian dynamics.

A striking case where such backflow can be studied in great detail is the class of collision-based mechanisms for the microscopic modelling of system-environment interaction (including correlated baths), from Markovian evolution [12–14] all the way to explicitly non-Markovian ones [15]. In such models, a quantum system $S$ collides sequentially with the elements of a multiarticle environment, whose constituents might or might not mutually interact. Each collision results in an inevitable pouring of information from the system into an environmental element. The resulting system’s dynamics is fundamentally dependent on whether or not such intra-environment interactions take place [12–15]. Suitable conditions (on the form and nature of the system-environment and environment-environment interaction and preparation) can be established that favor the emergence of strong non-Markovian features.

In this paper we dig into the analysis of the collisional model by studying the role played, in the establishment of overall non-Markovian features, by the correlations that are set between the system and the environmental element it has interacted with. In our analysis, we take the viewpoint provided by the trace norm-based measure of non-Markovianity [4], thus putting us in a perfect position to address the overall role played by SECs in the generation of non-Markovianity. By tracking the evolution of SECs as the system interacts with the elements of a multipartite environment, we are able to pinpoint a fundamental difference in the way system-environment interaction is modelled, which in turn results in quantitatively different manifestations of non-Markovianity. Our analysis is confirmed by a numerical “experiment”, implemented using quantum Monte
I. DESCRIPTION OF THE COLLISION MODEL

We consider the dynamics of a qubit system (labelled $S$) with logical states $\{|0\rangle, |1\rangle\}_S$ that interacts sequentially with the elements $\{E_1, E_2, ..., E_N\}$ of an N-party environment. The label that identifies each sub environment is provided by the position occupied by the corresponding environmental element in a spatial lattice. While the elements of the super-environment obviously be of any nature, for the sake of illustration here we focus on the case of simple two-level systems whose logical states are labelled as $\{|0\rangle, |1\rangle\}_j$ ($j = 1, ..., N$).

 Needless to say, as far as the interaction between $S$ and the $j^{th}$ element of the super-environment is concerned, many choices are available. Here we concentrate on the case of a coherent interaction (i.e. a mechanism that can be described by a Hamiltonian model of some form) giving rise to the unitary time-evolution operator

$$\hat{U}_{S,j}(\gamma) = \cos(\gamma) \mathbb{1}_{S,j} + i \sin(\gamma) \hat{S}_{S,j}.$$  

Here $\mathbb{1}_{S,j}$ is the identity operator, $\gamma \in \mathbb{R}$, is a dimensionless interaction strength for the interaction, and we have introduced the swap gate $\hat{S}_{S,j}$ that, in the ordered basis $\{|k, l\}_S\rangle$ (with $k, l = 0, 1$) reads [8]

$$\hat{S}_{S,j} = \begin{pmatrix}
1 & 0 & 0 & 0 \\
0 & 0 & 1 & 0 \\
0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1
\end{pmatrix}. \tag{2}$$

A similar model rules the interaction between two nearest-neighbor elements of the super-environment. That is, differently to the case addressed in Ref. [15] where a stochastic process regulates the subenvironment-subenvironment coupling, here we consider the unitary evolution

$$\hat{\mathcal{E}}_{j,j+1}(\delta) = \cos(\delta) \mathbb{1}_{j,j+1} + i \sin(\delta) \hat{S}_{j,j+1} \tag{3}$$

with $\delta \neq \gamma$, in general, and $\mathbb{1}_{j,j+1}, \hat{S}_{j,j+1}$ the analogue of the operations introduced above, defined in the Hilbert space of the elements $j$ and $j + 1$ of the super-environment. Such interactions give rise to the dynamical maps

$$\hat{\Phi}_{S,j}[\rho] = \hat{U}_{S,j}(\gamma) \rho \hat{U}_{S,j}^\dagger(\gamma),$$

$$\hat{\Psi}_{j,j+1}[\rho] = \hat{\mathcal{E}}_{j,j+1}(\delta) \rho \hat{\mathcal{E}}_{j,j+1}^\dagger(\delta). \tag{4}$$

Our model consists of sequential system-environment interactions interspersed with subenvironment-subenvironment couplings, according to the general scheme illustrated in Fig. 1: after the evolution induced on the joint state of $S$ and the $j^{th}$ subenvironment by $\hat{S}_{S,j}$, the system shifts by one site in the lattice while the $j^{th}$ environmental two-level system interacts with its rightmost nearest neighbour. Given the generic factorized initial state $\rho_0^{S,E}$, the overall evolution after interaction with $n$ elements of the super-environment can be formally written in terms of the unitary map

$$\rho_n^{S,E} = \hat{U}_n^{S,E} \hat{U}_n^\dagger$$  

with $\hat{U}_n$ the overall unitary evolution experienced by the $S$-$E$ system that is generated by the composition of the set of unitary gates introduced above. The dynamics of $S$ is then retrieved by discarding any information on the degrees of freedom of the super-environment, which breaks down the unitarity of the overall process. However, as we will discuss later on in this paper, the way such a process is accounted for is actually quite crucial and represents a very subtle point to consider.

In order to start making quantitative statements on the problem at hand, it is convenient to introduce one of the key instruments that will be used across this paper, namely the measure for non-Markovianity proposed in Ref. [4]. This is based on the study of the time-behavior...
of the trace distance between two generic states

$$D(\rho_1, \rho_2) = \frac{1}{2} \| \rho_1 - \rho_2 \|_1$$

(6)

with $\| \cdot \|_1$ the trace norm [8]. The trace distance is equal to 1 for fully distinguishable states and is null for identical states. If we take $\rho_{1,2}$ as the states of a system that is evolving under the action of a dynamical map starting from two different initial states, the degree of non-Markovianity is defined as

$$\mathcal{N} = \max_{\{\rho_1(0), \rho_2(0)\}} \int_{\Omega_+} \partial_t D(\rho_1(t), \rho_2(t)) dt$$

(7)

where $\Omega_+ = \bigcup_i (a_i, b_i)$ is the union of all the time intervals $(a_i, b_i)$ in our observation window within which $\partial_t D(\rho_1(t), \rho_2(t)) > 0$, $\rho_{1,2}(0)$ are two initial states of the system and $\rho_{1,2}(t)$ is their time-evolved form. The maximisation is performed over all possible pairs of initial states. The function $\partial_t D(\rho_1(t), \rho_2(t))$ encompasses the condition for revealing the non-Markovianity of an evolution: the existence of even a single region where $\partial_t D(\rho_1(t), \rho_2(t)) > 0$ is sufficient to guarantee non-Markovian nature of a dynamics. Conceptually, in fact, $\mathcal{N}$ accounts for all the temporal regions where the distance between two arbitrary input states increases, thus witnessing a re-flux of information from the environment to the system under scrutiny. Such a re-flux magnifies the difference between two arbitrarily picked input states evolved up to the same instant of time by the same map. Contractivity of the trace distance under divisible maps ensures that such re-flux never occurs, which in turn is equivalent to $\partial_t D(\rho_1(t), \rho_2(t)) \leq 0$. As the evolution under scrutiny here proceeds in discrete temporal steps, we will employ the discretised version of Eq. (7), which is obtained as

$$\mathcal{N} = \max \sum_n [D(\rho_{1,n}^S, \rho_{2,n}^S) - D(\rho_{n-1}^S, \rho_{2,n-1}^S)]$$

(8)

with $\rho_{k,n}^S$ the state of system $S$ obtained starting from the initial state $\rho_{k,0}^S$ after $n$ steps of our protocol. In the following, we will use the system preparation

$$\rho_k^S = \left( \begin{array}{cc} \cos^2 \theta_k & \cos \theta_k \sin \theta_k \\ \cos \theta_k \sin \theta_k & \sin^2 \theta_k \end{array} \right), \quad \theta_k \in [0, 2\pi]$$

(9)

standing for a pure initial state of $S$ determined by the angle $\theta_k$ in the Bloch sphere.

A. Quantum homogeneity

Armed with such a tool, we start noticing that, without introducing inter-ancilla collisions in our process, i.e. for $\delta = 0$, the model describes the process of quantum homogenization [12] whereupon preparation of an identical fiducial state for all the elements of the super-environment, the state of the system eventually homogenizes to it, thus realizing a microscopic model for Markovian decoherence. In order to set a useful benchmark for the comparison with quantum homogenisation, in our assessment of the non-Markovian features arising from the inclusion of intra-environment interactions we will also consider an identical fiducial state for the super-environment. In particular, we will consider the initialisation $|0\rangle \equiv \otimes_{j=1}^{S} |0\rangle_j$. In Fig. 2 we show the behaviour of trace distance $D$ and state fidelity $F = S(0)/\rho_{n}^S(0)$ between the state of the system after $n$ collisions with the super-environment and the target state $|0\rangle_S$ (i.e. the state into which each element of the environment is prepared). The trace decreases monotonically with the number of system-environment collisions, thus witnessing the complete absence of any backflow mechanisms that might give rise to non-Markovian features. These conclusions hold qualitatively regardless of the initial preparation of the super-environment.

B. Description of the strategies

We are now in a position to attack the main goal of this paper and address the delicate point of tracking the reduced dynamics of the system $S$. We will focus on two inequivalent ways of tracing out the degrees of freedom of the super-environment. In turn, this will allow us to pinpoint the key role played by SECs in the settling of non-Markovian features.

The first method that we use in order to compute the reduced dynamics of $S$ (which we dub Strategy 1) is that of tracing out one of the subenvironments as soon as the system has interacted with it. This corresponds to taking a “utilitarian” viewpoint according to which element $E_j$ of the super-environment is relevant to determining the evolution of $S$ only as far as their mutual interaction is concerned. This means that correlations between the $S$ and $E_j$ are erased before the interaction between $E_j$ and $E_{j+1}$ occurs and cannot, therefore, affect $S$ during the next iteration of interactions. In this respect, the effect...
that the “collision” with $S$ has on the state of $E_j$ is carried over across the super-environment regardless of the actual state of $S$. This is, that the reduced state of the system qubit after the interaction with a set of $n$ subenvironments is described by the dynamical map

$$\rho_n^S = \text{Tr}_{n-1,n}(\tilde{\psi}_{n-1,n}[\tilde{\psi}_{n,n}(\rho_{n}^{SE} \otimes |0\rangle\langle0|)]),$$

with $\text{Tr}_{S}(\rho_{n-1}^{SE})$ the reduced state of the $(n-1)$th element of the subenvironment after its interaction with the system qubit, which is in turn left in state $\rho_n^S$.

The second approach (named Strategy 2 hereafter) implies the tracing out of a subenvironment only after it has allowed its usefulness. In more explicit terms, we consider the various subenvironments in a time-non-local fashion: element $E_j$ will be traced out only after its active role in the collision model has expired, i.e. when it has interacted with the ordered triplet $(E_{j-1}, S, E_{j+1})$.

The dynamical map resulting from the implementation of this strategy thus gives rise to the reduced state of the system

$$\rho_n^S = \text{Tr}_{n-1,n}(\tilde{\psi}_{n-1,n}[\tilde{\psi}_{S,n}(\rho_{n}^{SE} \otimes |0\rangle\langle0|)]),$$

where $\text{Tr}_{(n-2)}[\cdot]$ denotes the partial trace over the whole set of $n-2$ elements of the super-environment prior to the interaction between $S$ and element $E_{n-1}$. Quite intuitively, the difference between the two strategies resides in the different way SECs are treated: while Strategy 1 erases all the SECs established by a given system-subenvironment collision, the second one carries these over to the next intra-environment interaction. This results in considerable differences in the non-Markovianity features arising from the dynamical maps formalised by Eq. (10) and (11). However, contrarily to a naive expectation, Strategy 1 does not give rise to a homogenisation process like the one addressed earlier on, in light of a non negligible environmental memory effect, and leaves room for non-Markovian manifestations. We have checked that the qualitative features that will be showcased throughout our analysis depend critically on the degree of purity of the overall super-environmental state, but only weakly on its explicit form.

II. ANALYSIS OF THE DEGREE OF NON-MARKOVIANITY

Here we present our analysis of the non-Markovian features of the collision model addressing both the strategies identified above.

A. Non-Markovianity resulting from both Strategies

We start analysing the behaviour of the trace distance $D(\rho_0, \rho_n)$ as the collision-based model for system-environment interaction is iterated. When using Strategy 2 with $\delta = \pi/2$ (i.e. for a full state-swap between two consecutive super-environmental elements), the joint dynamics of $S$ and $E$ resembles that of an iterated two-qubit system. This is due to the complete exchange of information between environment qubits at every step, which makes the system interact at step $n$ with a fresh physical information carriers that, however, carries fully the effect of the collision occurred at step $n-1$. This results, needless to say, in dynamics characterised by undamped oscillations of the trace distance, which would in turn give rise to a degree of non-Markovianity that would grow with the size of the temporal window of observation of the evolution reaching, asymptotically, an infinite value [cf. Fig. 3(a)].

However, requiring $\delta = \pi/2$ implies, in quite a general sense, a strong intra-environment interaction. Unsurprisingly, this results in the non-Markovian features highlighted above, in light of the pronounced dynamical nature of the corresponding super-environment. It is thus interesting to address the case of $\delta < \pi/2$, i.e. a weaker subenvironment-subenvironment coupling strength. The expectation is that this would correspond to a loss of information over the state of the $n^{th}$ environmental element, whose state is only partially carried over to element $(n + 1)^{th}$. This is well captured by the trace distance, which oscillates with a degraded amplitude, as illustrated in Fig. 3(b).

The trend shown in Figs. 3(c) and (d) highlight the differences between the two strategies addressed in our study. In fact, when Strategy 1 is employed to model the

FIG. 3: The trace distance $D$ plotted against the number of iterations of the collision model for both the strategies and for different swap strengths put in place in our analysis. Panels (a) and (b) [(c) and (d)] report the results valid for Strategy 2 [Strategy 1]. We have used $\delta = \pi/2$ in panels (a) and (c) and $\delta = 0.95 \times \pi/2$ in panels (b) and (d) to show how varying the inter-ancilla swap strength changes substantially the degree of non-Markovianity. We have taken $\gamma = 0.05$ for the system-environment interaction strength, consistently with the assumption of weak S-E coupling.
place the measure stated in Eq. (7) for different values of the difference between the two schemes by putting in (the pair interaction strength every iteration in both system-environment and intra-system are inputted. The second term deals with the effects due to non-null SECs. We have examined the respective contributions from the two terms to explore the origin of the non-Markovianity we observed in Section II A. The corresponding results are shown in Fig. 5: The upper bound is found to be completely formed from SECs, the term corresponding to environmental differences being null when the input system states are mutually orthogonal. In turn, this explains why the derivative of the trace distance corresponding to Strategy 1 was maximised for values other than the |0⟩ and the |1⟩ state. Indeed, as the origin of non-Markovianity relies entirely on the establishment of SECs, orthogonal input states in the computational basis in Strategy 1, which cancels all of them, would only give rise to N = 0. It is also worth mentioning that when Eq. (12) is computed using Strategy 2 and the input states that are optimal for Strategy 1, we do observe a contribution to the quantitative value of the upper bound to β(t, ρ^S_0) coming from the dynamical nature of the environment [i.e. the first term in Eq. (12)]. Such contribution becomes irrelevant for the optimal case of orthogonal input S states.

B. The role of SECs

S-E interaction, even the strongest intra-environmental coupling strength produces a depleted back-flow mechanism: the initially significant oscillations of the trace distance gradually fade to a small yet non-null value. While the dynamics persists to be non-Markovian even under Strategy 1, the qualitative features of the evolution are indeed strongly dependent on the way information is propagated across. As we will argue in the following Section, the reasons behind the differences between the two dynamical strategies are due to the different way SECs are accounted for. Also, while for the case of Strategy 2 the maximum inherent in the definition of the measure of non-Markovianity used here is achieved for the input states |0⟩ and |1⟩, for Strategy 1 we require the pair (|0⟩ − |1⟩)/√2, (|0⟩ + |1⟩)/√2. We can quantify the difference between the two schemes by putting in place the measure stated in Eq. (7) for different values of δ. In Fig. 4 the results associated with the two schemes are shown for 100 different values of δ ∈ [0, π/2]. Strategy 2 is spectacularly superior to Strategy 1 in setting non-zero degree of non-Markovianity even at moderate values of the S-E interaction and has a comparably smaller threshold in the value of δ above which the dynamics is signalled as non-Markovian. For δ = γ = π/2, N = n. In this case, in fact, we have complete swaps at every iteration in both system-environment and intra-environment interactions.

FIG. 4: Measure N plotted against the intra-environment interaction strength δ, for both the strategies addressed in this work. We have used γ = 0.05, n = 3 × 10^5, and the super-environment is initialised in |{0}⟩. By approaching the full-swap condition embodied by δ = π/2, the non-Markovianity measure shoots up. The vertical axis of the plot has been truncated to N = 10 so as to improve the visibility of the details of the plot. The qualitative and quantitative differences inherent in the different strategies for the modelling of S-E interactions are evident in the different thresholds in the value of δ above which N > 0. The measure is optimised over all possible pairs of initial S states.

so as to improve the visibility of the details of the plot. The qualitative and quantitative differences inherent in the different strategies for the modelling of S-E interactions are evident in the different thresholds in the value of δ above which N > 0. The measure is optimised over all possible pairs of initial S states.
We have extended our numerical analysis by considering aleatory system-environment interactions: in the spirit of a Monte Carlo simulations, we have introduced a random variable in our iterative model: should such variable take a value smaller than a chosen threshold (which we let span the range of values $[0, 1]$), $S$ and $E_j$ would interact at step $j$ of the evolution. This process was examined for varying threshold values in Fig. 6. Clearly, as the threshold is increased we allow for system environment interactions to occur, thus increasing the resulting degree of non-Markovianity, which appears to depend linearly on the chosen threshold.

We have found that, for $\delta = \pi/2$, such numerical experiment yields changes only in the actual degree of non-Markovianity, which depends on the value taken by the threshold. This is due to the fact that the full swap occurring at the sub-environmental level is not at all affected by a “missed” system-environment collision: such an event would merely shift the feedback of the environment into the system to the next “allowed” interaction. Decreasing the probability of $S$-$E$ interaction will only affect the period of the oscillations of the trace distance, leaving their amplitude unaffected.

III. CONCLUSIONS AND OUTLOOK

We have studied the non-Markovian phenomenology arising from a collision-based microscopic model for system-environment interaction. Our analysis focused on the role that SECs play in the settling of non-Markovian features in the system’s dynamics: by putting in place recently proposed tools for the in-depth analysis of the trace distance-based measure of non-Markovianity, and addressing explicitly two non-equivalent iterative protocols for the joint evolution of the system and a multi-particle environment, we have been able to provide evidences of the actual crucial contribution of SECs for the determination of the actual degree of non-Markovianity and the characterisation of the details of such evolution. A Monte Carlo-inspired numerical modelling, built by biasing the chance that a given system-environment interaction actually occurs, showed the persistence of the non-Markovian character of the overall evolution. This analysis opens up interesting avenues for the from thermodynamic-inspired exploration of non-Markovianity in collision-based models.

Acknowledgments

We thank F. Ciccarello for his helpful insight in the development of this project, Laura Mazzola for her invaluable discussions and André Xuereb for his constructive suggestions. RMCC acknowledges financial support from the Northern Ireland DEL. MP thanks the UK EPSRC for a Career Acceleration Fellowship and a grant awarded under the “New Directions for Research Leaders” initiative (EP/G004579/1), the Alexander von Humboldt Stiftung, the John Templeton Foundation (grant 43467), and the EU project TherMiQ (grant agreement 618074).


