Tensor-Train Thermo-Field Memory Kernels for Generalized Quantum Master Equations

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ABSTRACT: The generalized quantum master equation (GQME) approach provides a rigorous framework for deriving the exact equation of motion for any subset of electronic reduced density matrix elements (e.g., the diagonal elements). In the context of electronic dynamics, the memory kernel and inhomogeneous term of the GQME introduce the implicit coupling to nuclear motion and dynamics of electronic density matrix elements that are projected out (e.g., the off-diagonal elements), allowing for efficient quantum dynamics simulations. Here, we focus on benchmark quantum simulations of electronic dynamics in a spin-boson model system described by various types of GQMEs. Exact memory kernels and inhomogeneous terms are obtained from short-time quantum-mechanically exact tensor-train thermo-field dynamics (TT-TFD) simulations and are compared with those obtained from an approximate linearized semiclassical method, allowing for assessment of the accuracy of these approximate memory kernels and inhomogeneous terms. Moreover, we have analyzed the computational cost of the full and reduced-dimensionality GQMEs. The scaling of the computational cost is dependent on several factors, sometimes with opposite scaling trends. The TT-TFD memory kernels can provide insights on the main sources of inaccuracies of GQME approaches when combined with approximate input methods and pave the road for the development of quantum circuits that implement GQMEs on digital quantum computers.

1. INTRODUCTION

Quantum dynamics simulations are central to theoretical studies of many areas of chemical and technological applications, including charge and energy transfer in photosynthetic and photovoltaic systems and a wide range of reactions with nonadiabatic dynamics and photochemical processes, including spin and vibrational energy relaxation, as well as polaritonic chemistry. Despite considerable progress over the past few decades, the development of efficient methods for simulations of quantum dynamics remains an outstanding challenge for studies of complex molecular systems at finite temperature. This is primarily due to the computational cost of quantum-mechanically exact simulations, which scales exponentially with the number of degrees of freedom in the system, thereby making such simulations intractable in most complex molecular systems of practical interest. Thus, reduced-dimensionality approaches that can offer more favorable scaling are highly desirable.

The Nakajima–Zwanzig generalized quantum master equation (GQME) provides a formally exact general-purpose framework for modeling quantum dynamics in reduced dimensionality. It can be obtained for any subset of reduced density matrix elements by using suitable projection operators. When focusing on electronic dynamics, the effect of projecting out nuclear degrees of freedoms (DOFs) and electronic density matrix elements not included in the subset of interest is accounted for by the memory kernel and the inhomogeneous term of the GQME. The dimensionality of those spatially and temporally compact quantities is typically much lower than the dimensionality of the overall system since it is determined by the number of reduced density matrix elements included in the subset of interest, allowing for efficient simulations.

Considerable progress has already been made toward calculating the aforementioned memory kernels and inhomogeneous terms without resorting to perturbation theory. Much of that progress has been based on the strategy introduced by Shi and Geva, which relies on formally exact relationships between the memory kernel and the inhomogeneous term and projection-free inputs (PFIs) that are given in terms of two-time correlation functions of the overall system. These PFIs can be obtained from quantum-mechanically exact or approximate (e.g., semiclassical or mixed quantum-classical) input methods.

In this paper, we introduce exact memory kernels and inhomogeneous terms obtained from quantum-mechanically exact tensor-train thermo-field dynamics (TT-TFD) simulations. To the best of our knowledge, this is the first

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application of TT-TFD to calculate memory kernels and inhomogeneous terms of GQMEs. Previously, exact memory kernels have been obtained by the Gevi,27 Shi,37–50 Makri,51 and Rabani31,33,34,36,55 groups. This paper extends the available exact results to include the memory kernels and inhomogeneous terms of the modified GQME and reduced-dimensionality GQME approaches for the spin-boson model. We demonstrate the capabilities of the GQMEs as applied to benchmark simulations of electronic relaxation dynamics in a spin-boson model system, including calculations based on various types of reduced-dimensionality GQMEs. The spin-boson model provides a useful framework for modeling molecular systems with coupled electronic states. The resulting quantum-mechanically exact memory kernels and inhomogeneous terms can serve as benchmarks for assessing the accuracy of approximate memory kernels and inhomogeneous terms obtained by approximate input methods. In addition, the reported quantum-mechanically exact memory kernels and inhomogeneous terms enables the development of quantum circuits for the implementation of GQMEs on digital quantum computers, which we report in the upcoming work.56

The paper is organized as follows. The objectives and scope of our approach are presented in Section 2, the GQME formalism is outlined in Section 3, and the protocol used for calculating the PFIs via TT-TFD is described in Section 4. The utility of combining the GQME and TT-TFD approaches is demonstrated for the benchmark spin-boson model in Section 5. Also included in Section 5 is a comprehensive comparison between the TT-TFD-based quantum-mechanically exact results and the corresponding approximate results based on PFIs obtained with an approximate linearized semiclassical mapping Hamiltonian method.19 Concluding remarks are provided in Section 6. Additional graphs and computational details are included in the Supporting Information (SI).

2. MODEL SYSTEM

We focus on molecular systems exhibiting nonadiabatic quantum dynamics such as photosynthetic and photovoltaic molecular assemblies, commonly described by the following model Hamiltonian:

\[
\hat{H} = \sum_{j=1}^{N_{e}} \hat{H}_{j}|j\rangle\langle j| + \sum_{j,k=1}^{N_{f}} \hat{V}_{jk}|j\rangle\langle k|
\]

Here, \(\hat{H}_{j}\) is the nuclear Hamiltonian when the system is in diabatic electronic state \(|j\rangle\) \([\hat{H}_{j} = p_{j}^{2}/2 + V_{j}(\hat{R})]\), with index \(j\) running over the \(N_{e}\) electronic states \((j = 1, 2, \ldots, N_{e})\), while \(\hat{R} = (\hat{R}_{1}, \ldots, \hat{R}_{N_{f}})\) and \(\hat{P} = (\hat{P}_{1}, \ldots, \hat{P}_{N_{f}})\) are the mass-weighted position and momentum operators of the \(N_{f} \gg 1\) nuclear DOF, and \(\{\hat{V}_{jk}\} \neq k\) are coupling terms between electronic states which can be either nuclear operators (non-Condon case) or constants (Condon case). Throughout this paper, a circumflex symbol over a variable (e.g., \(\hat{R}\)) indicates an operator quantity and a script font (e.g., \(\mathcal{L}\)) indicates a superoperator.

For simplicity, we assume that the initial state of the overall system has the single-product form,

\[
\hat{\rho}(0) = \hat{\rho}_{n}(0) \otimes \hat{\sigma}(0)
\]

Here, \(\hat{\rho}_{n}(0) = \text{Tr}_{e}\{\hat{\rho}(0)\}\) and \(\hat{\sigma}(0) = \text{Tr}_{n}\{\hat{\rho}(0)\}\) are the reduced density operators that describe the initial states of nuclear DOF and electronic DOF, respectively, while \(\text{Tr}_{e}\{\cdot\}\) and \(\text{Tr}_{n}\{\cdot\}\) represent partial traces over the electronic and nuclear Hilbert spaces, respectively. The methodology presented in this paper is not limited to factorized initial states, as introduced by eq 2, and can be applied to arbitrary initial states.55

The time-dependent propagation of the initial state, introduced by eq 2, according to the Hamiltonian introduced by eq 1, yields the propagated state \(\hat{\rho}(t)\) at time \(t\), which is described by the following density operator:

\[
\hat{\rho}(t) = e^{-i\hat{H}/\hbar} \hat{\rho}_{n}(0) \otimes \hat{\sigma}(0) e^{i\hat{H}/\hbar} \equiv e^{-\mathcal{L}(\cdot)\hbar/2} \hat{\rho}_{n}(0) \otimes \hat{\sigma}(0)
\]

(3)

Here, \(\mathcal{L}(\cdot)\) is the overall Liouvillian superoperator \((\mathcal{L}(\cdot) = [\hat{H}, \cdot])\). The reduced electronic density operator \(\hat{\sigma}(t)\) at time \(t\) is obtained by tracing out the nuclear, as follows:

\[
\hat{\sigma}(t) = \text{Tr}_{n}\{\hat{\rho}(t)\} = \sum_{j,k=1}^{N} \sigma_{jk}(t)|j\rangle\langle k|
\]

(4)

The electronic populations and coherences are given by \(\{\sigma_{jk}(t) = (|j\rangle\langle k|)\}\) and \(\{\sigma_{k}(t) = (|j\rangle\langle k|)|j \neq k\}\), respectively. These quantities are of particular interest, because their time evolution underlies electronic energy, charge, and coherence transfer dynamics, as well as electronic decoherence.

3. GQMEs in Reduced Dimensionality

The GQME formalism can be applied to derive exact equations of motion for electronic observables while keeping the input regarding other DOFs in the system to the minimum necessary to account for their impact on dynamics. To this end, we begin with the well-known Nakajima–Zwanzig GQME (whose derivation is outlined in the SI),

\[
\frac{d}{dt}\mathcal{P}\hat{\rho}(t) = -\frac{i}{\hbar}\mathcal{P}\mathcal{L}\hat{\rho}(t) - \frac{1}{\hbar} \int_{0}^{t} d\tau \mathcal{P}Q_{c}(\tau)\mathcal{L}\hat{\rho}(t - \tau) - \frac{i}{\hbar}\mathcal{P}Q_{c}(t)\mathcal{P}\hat{\rho}(0)
\]

(5)

where \(\mathcal{P}\) is a projection superoperator.17,19 Here, \(Q\) is the projection superoperator complementary to \(\mathcal{P}\) \((Q = I - \mathcal{P})\), \(I\) is the identity superoperator, and \(\mathcal{L}\) is the Liouvillian superoperator as in eq 3. Integrating eq 5, we obtain the time-dependent projected state \(\mathcal{P}\hat{\rho}(t)\). Importantly, there is a lot of flexibility in the choice of \(\mathcal{P}\) to select the specific quantity of interest.19

In this paper, we focus on quantities of interest corresponding to a subset of electronic reduced density matrix elements \(\{\sigma_{jk}(t)\}\). For example, \(\{\sigma_{0j}(t)\}\) may include all \(N_{e}^{2}\) electronic reduced density matrix elements (i.e., all populations and coherences), in which case \(\{\sigma_{jk}(t)\} = \{\sigma_{0j}(t), \ldots, \sigma_{\sqrt{N_{e}}N_{f}}(t), \ldots, \sigma_{N_{f}N_{e}}(t)\}\) or only the diagonal electronic reduced density matrix elements (i.e., the populations of the corresponding electronic states), in which case \(\{\sigma_{0i}(t)\} = \{\sigma_{ii}(t), \ldots, \sigma_{N_{f}N_{f}}(t)\}\) or just the single diagonal term describing the time-dependent population of state \(\mid i\rangle\), in which case \(\{\sigma_{0i}(t)\} = \{\sigma_{ii}(t)\}\). The GQME of an element \(\sigma_{0i}(t)\) in a specific subset of electronic reduced density matrix elements \(\{\sigma_{0i}(t)\}\) is defined as
\[
\frac{d}{dt} \sigma_n(t) = -\frac{i}{\hbar} \sum_{m \in \{ab\}} \langle \mathcal{L}_{jk,lm} \rangle^0_n \sigma_m(t) \\
- \sum_{m \in \{ab\}} \int_0^t d\tau \kappa_{jk,lm}(\tau) \sigma_m(t - \tau) + \hat{I}_{jk}^n(t)
\]

Equation 6 is different for different subsets of electronic reduced density matrix elements. It can be obtained by first defining the projection superoperator

\[
\mathcal{P}_{\text{set}} \hat{A} = \sum_{jk \in \{ab\}} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \hat{A} \langle 0 \rangle \otimes \langle jk \rangle
\]

where \( \hat{A} \) is a general overall system operator, \{ab\} are the indices matching the subset of electronic reduced density matrix elements \( \langle \sigma_{ab}(t) \rangle \), and \( \hat{I}_n \) is the unity operator in the nuclear Hilbert space. Equation 6 can then be obtained by substituting the projection superoperator \( \mathcal{P}_{\text{set}} \) defined in eq 7 into eq 5, tracing over the nuclear Hilbert spaces, and applying \( \langle jk \rangle \) from the left and \( \hat{I}_n \) from the right, with a detailed derivation included in the SI (section S1.2). Note that, because the sum in eq 7 goes over the elements of the subset of interest, \( \mathcal{P}_{\text{set}} \) will differ for different subsets and, therefore, the equation of motion in eq 6 will also be different, even for the same element when it is in different subsets. For example, if we are looking at one subset \( \{ \sigma_{ab}(t) \} = \{ \sigma_{11}(t) \} \) and another subset \( \{ \sigma_{ab}(t) \} = \{ \sigma_{12}(t), \sigma_{21}(t) \} \), the equation of motion for \( \sigma_{11}(t) \) given in eq 6 will be different for the first subset versus the second subset. These different equations of motion are exact, so if exact input methods are used to obtain the terms on the right-hand side (RHS) of eq 6 for \( \sigma_{11}(t) \), the results will be the same. However, if approximate input methods are used, the results can differ.\(^{19} \)

In eq 6, \( \langle \mathcal{L}_{jk,lm} \rangle^0_n \), \( \kappa_{jk,lm}(\tau) \), and \( \hat{I}_{jk}^n(t) \) are the matrix elements \( (jk, lm) \) of the projected Liouvilian superoperator, memory kernel superoperator, and inhomogeneous term operator, respectively, which are defined as follows:

\[
\langle \mathcal{L}_{jk,lm} \rangle^0_n = \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L}_n \langle 0 \rangle \otimes \langle ml \rangle
\]

\[
\kappa_{jk,lm}(\tau) = \frac{1}{\hbar} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \rho_{ml}(0) \otimes \langle ml \rangle
\]

\[
\hat{I}_{jk}^n(t) = -\frac{i}{\hbar} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \hat{\rho}(0) - \sum_{m \in \{ab\}} \hat{\rho}_m(0) \otimes \langle ml \rangle \langle ml \rangle \sigma_m(0) \}
\]

These quantities are subset-dependent, so they will differ depending on the chosen subset of electronic density matrix elements of interest. Given that \( N_{\text{set}} \) is the number of matrix elements of interest included in \( \{ \sigma_{ab}(t) \} \) (1 \( \leq N_{\text{set}} \leq N_N^2 \)), the projected Liouvilian \( \langle \mathcal{L} \rangle^0 \) and memory kernel \( \kappa_{\text{set}}(\tau) \) superoperators can be represented by \( N_{\text{set}} \times N_{\text{set}} \) matrices, whereas the inhomogeneous term operator \( \hat{I}_{\text{set}}^n(t) \) can be represented by an \( N_{\text{set}} \)-dimensional vector in Liouville space.

Calculating the projected Liouvillian is typically straightforward. The memory kernel and inhomogeneous term satisfy Volterra integral equations, so they can be obtained from the PFIs.\(^{19} \) The Volterra equation for the memory kernel is given by

\[
\kappa_{jk,lm}(\tau) = i\mathcal{F}_{jk,lm}(\tau) - \frac{1}{\hbar} \sum_{uv \in \{ab\}} \mathcal{F}_{jk,uv}(\tau) \langle \mathcal{L}_{uv,lm} \rangle^0_n
\]

\[
+ i \sum_{uv \in \{ab\}} \int_0^\tau d\tau' \mathcal{F}_{jk,uv}(\tau - \tau') \kappa_{uv,lm}(\tau')
\]

where the PFIs are given by

\[
\mathcal{F}_{jk,lm}(\tau) = \frac{1}{\hbar} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \hat{\rho}(0) \otimes \langle ml \rangle
\]

\[
\mathcal{F}_{jk,lm}(\tau) = -\frac{1}{\hbar} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \rho_{ml}(0) \otimes \langle ml \rangle
\]

The Volterra equation for the inhomogeneous term is given by

\[
\hat{I}_{jk}^n(t) = \mathcal{Z}_{jk}(t) + i \sum_{m \in \{ab\}} \mathcal{F}_{jk,lm}(t) \sigma_m(0)
\]

\[
+ i \sum_{uv \in \{ab\}} \int_0^t d\tau' \mathcal{F}_{jk,uv}(t - \tau') \hat{I}_{uv}^m(\tau')
\]

where the additional PFI \( \mathcal{Z}_{jk}(t) \) is given by

\[
\mathcal{Z}_{jk}(t) = -\frac{i}{\hbar} \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} \mathcal{L} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \hat{\rho}(0)
\]

Note that \( \mathcal{Z}_{jk}(t) = -i\mathcal{F}_{jk,lm}(t) \) when the overall initial state is of the commonly encountered form \( \hat{\rho}(0) = \hat{\rho}(0) \otimes \hat{l}_a \) (where \( \hat{l}_a \) is one of the electronic basis states), as is the case for the applications reported in this paper. A more-detailed discussion of the derivation, properties, and significance of eqs 11–14 can be found in ref 19 and in the SI.

Most previous studies have been based on direct calculations of the aforementioned PFIs,\(^{19,43,43–46} \) However, when using an exact input method, the PFIs can also be accurately obtained as derivatives of the propagator \( \mathcal{U}(\tau) \equiv \text{Tr}_n[\mathcal{L}] e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \hat{\rho}(0) \otimes \hat{l}_a \) that evolves the electronic reduced density operator, as follows:\(^{41,43} \)

\[
\hat{\sigma}(\tau) = \mathcal{U}(\tau) \hat{\sigma}(0)
\]

with matrix elements,

\[
\mathcal{U}_{jk,lm}(\tau) = \text{Tr} \{ \langle jk \rangle \otimes \hat{I}_n \} e^{-\mathcal{L}^{T T} \mathcal{L} / \hbar} \hat{\rho}(0) \otimes \langle ml \rangle
\]

Specifically, we obtain the PFIs \{ \( \mathcal{F}\}_{jk,lm}(\tau), \mathcal{F}_{jk,lm}(\tau) \} from \{ \( \mathcal{U}\}_{jk,lm}(\tau) \}, as follows:\(^{41,43} \)

\[
\mathcal{F}_{jk,lm}(\tau) = i \mathcal{U}_{\text{FF},lm}(\tau) \quad \mathcal{F}_{jk,lm}(\tau) = i \mathcal{U}_{\text{FF},lm}(\tau)
\]

PFIs \{ \( \mathcal{F}\}_{jk,lm}(\tau), \mathcal{F}_{jk,lm}(\tau) \} calculated from \( \mathcal{U}(\tau) \) generate exact memory kernels and inhomogeneous terms when \( \mathcal{U}(\tau) \) is obtained from exact inputs. Therefore, we can calculate them in terms of numerical derivatives of \( \mathcal{U}(\tau) \) obtained from quantum-mechanically exact TT-TFD simulations, as described in Section 4.
4. TENSOR-TRAIN THERMO-FIELD DYNAMICS

4.1. Hamiltonian. The molecular Hamiltonian introduced by eq 1 can also be written as the sum of a purely electronic Hamiltonian $\hat{H}_e \otimes \hat{I}_n$ plus a purely nuclear Hamiltonian $\hat{I}_n \otimes \hat{H}_{n\prime}$, and an interaction term between the electronic and nuclear DOF, $\hat{H}_{iG}$:

$$\hat{H} = \hat{H}_e \otimes \hat{I}_n + \hat{I}_n \otimes \hat{H}_{n\prime} + \hat{H}_{iG}$$

(18)

Note that this division is not unique, in the sense that different choices of $\hat{H}_e$, $\hat{H}_{n\prime}$, and $\hat{H}_{iG}$ are possible. However, the results are invariant to those choices when a quantum-mechanically exact method like TT-TFD is applied, since no physical or ad hoc approximation is introduced.

4.2. Thermo-Field Dynamics Method. We start out by noting that the dynamics of $\hat{\rho}(t)$, which is governed by a Hamiltonian of the form of eq 18, is described by the quantum Liouville equation,

$$\frac{d}{dt}\hat{\rho}(t) = -\frac{i}{\hbar} [\hat{H}, \hat{\rho}(t)]$$

(19)

The TT-TFD method provides a generally numerically exact approach to solve eq 19 that is particularly efficient when $\hat{\rho}(t)$ can be represented as a low rank matrix product state. In our simulations, the state is described by $\hat{\rho}^{1/2}(t)$ [instead of $\hat{\rho}(t)$], represented as a tensor-train vector in an extended Hilbert space (the so-called double Hilbert space described below). The Liouville equation given in eq 19 is replaced by an equivalent equation of motion for $\hat{\rho}^{1/2}(t)$, which can be written in the form of a Schrödinger-like equation in the double Hilbert space. For a high-dimensional system, computational efficiency is achieved by using a tensor-train representation of the extended state vector $\hat{\rho}^{1/2}(t)$. The remainder of this section outlines the TT-TFD methodology used for calculating the PFIs needed to obtain the memory kernel and inhomogeneous term of the GQMEs.

The initial density operator of the overall system is of the form introduced by eq 2. The initial electronic density operator is given by $\hat{\rho}(0) = |\gamma\rangle\langle\gamma|$, where $|\gamma\rangle$ is one of the electronic basis states, while the initial nuclear density operator is $\hat{\rho}_{n\prime}(0) = e^{-\hat{H}_{n\prime}}/Z_n(\beta)$, where $Z_n(\beta) = \text{Tr}_e(e^{-\hat{H}_{n\prime}})$. Therefore,

$$\hat{\rho}(0) = |\gamma\rangle\langle\gamma| \otimes e^{-\hat{H}_{n\prime}}/Z_n(\beta)$$

(20)

However, we note that the TT-TFD method is not restricted to initial states of this simple form and can be analogously applied to propagate any arbitrary initial state.

The TFD representation is only applied to the nuclear density operator of the system, since the same dynamics is obtained for the initial state introduced by eq 20, regardless of whether the electronic density operator is included or not in the TFD representation. We let $\{|k\rangle\}$ be an orthonormal basis that spans the physical nuclear Hilbert space $\mathcal{H}_n$, and $\{|k\rangle\}$ be an orthonormal basis that spans a fictitious nuclear Hilbert space (also known as the tilde space) $\mathcal{H}_{n\prime}$, which is an exact replica of $\mathcal{H}_n$. Next, we define the so-called nuclear thermal vacuum state:

$$|0_n(\beta)\rangle = e^{-\hat{H}_{n\prime}/2} \sum_{k=1}^{\infty} |k\rangle \otimes |k\rangle$$

(21)

where one should note that the sum includes only terms $|k\rangle \otimes |k\rangle$ with $i t h k = k$, $s o t h a t \sum_{k=1}^{\infty} |k\rangle \otimes |k\rangle = |0\rangle \otimes |0\rangle + |1\rangle \otimes |1\rangle + ...$. We note that $\hat{\rho}_{n\prime}(0)$ can be obtained from $|0_n(\beta)\rangle$, upon taking the outer product with its dual and tracing out the fictional degrees of freedom as follows:

$$\text{Tr}_f(|0_n(\beta)\rangle\langle0_n(\beta)|) = \hat{\rho}_{n\prime}(0)$$

(22)

where $\text{Tr}_f(\cdot)$ is the partial trace over states $|k\rangle$ in the tilde space $\mathcal{H}_{n\prime}$.

Substituting eq 22 into eq 20, we obtain the initial density operator of the overall system $\hat{\rho}(0)$ represented in terms of the ket vector $|\psi_n(\beta, 0)\rangle \equiv |\gamma\rangle \otimes |0_n(\beta)\rangle$, as follows:

$$\hat{\rho}(0) = \text{Tr}_f(|\psi_n(\beta, 0)\rangle\langle\psi_n(\beta, 0)|)$$

(23)

Note that, in eq 23, only the initially thermalized nuclear density operator is represented by a ket vector in the double space $\mathcal{H}_n \otimes \mathcal{H}_{n\prime}$ whereas the initial electronic density operator $|\gamma\rangle\langle\gamma|$ corresponds to a pure state in the electronic Hilbert space.

We define the overall system ket vector $|\psi_n(\beta, t)\rangle$, such that

$$\hat{\rho}(t) = \text{Tr}_f(|\psi_n(\beta, t)\rangle\langle\psi_n(\beta, t)|)$$

(24)

where $\hat{\rho}(t)$ evolves according to the Liouville equation (eq 19). This can be fulfilled by evolving $|\psi_n(\beta, t)\rangle$ according to the so-called TFD Schrödinger equation (as shown in the SI),

$$\frac{d}{dt}|\psi_n(\beta, t)\rangle = -\frac{i}{\hbar} \Pi |\psi_n(\beta, t)\rangle$$

(25)

where $\Pi = \hat{H}_e \otimes \hat{I}_n$ with $\hat{I}_n = \sum_i |k\rangle\langle k|$ being the identity operator of the tilde space. Moreover, we note that the same physical system dynamics can be obtained by defining $\hat{H}$ in eq 25, as follows:

$$\Pi = \hat{H}_e \otimes \hat{I}_n - \hat{I}_n \otimes \hat{H}_{n\prime}$$

(26)

where $\hat{I}_n = \sum_i |k\rangle\langle k|$. Remarkably, $\hat{H}_e$ can be any operator in the nuclear tilde space since $\hat{H}_{n\prime}$ does not impact kets in the physical space and its effect on the dynamics vanishes upon taking the partial trace over states in the tilde space.

The preparation of the initial thermal wavepacket $|\psi_n(\beta, 0)\rangle$, according to eqs 21 and 22, requires the explicit evaluation of the quantum Boltzmann operator, which can be computationally challenging for systems with high dimensionality. However, when the initial nuclear Hamiltonian is harmonic, the initial thermal wavepacket can be obtained by taking advantage of the thermal Bogoliubov transformation. Therefore, we can generate the nuclear thermal vacuum state from the double space ground state $|0_n, 0_{n\prime}\rangle$ using the following unitary transformation:

$$|\psi_n(\beta, 0)\rangle = |\gamma\rangle \otimes e^{-\hat{G}/2}|0_n, 0_{n\prime}\rangle$$

(27)

where $\hat{G}$ is given by

$$\hat{G} = -i \sum_j (\hat{a}_j \hat{a}_j^\dagger - \hat{a}_j^\dagger \hat{a}_j)$$

(28)
with \( \theta = \text{arctanh}(e^{-\beta_0/2}) \), where \( \{a_i, a_i^\dagger\} \) and \( \{d_i, d_i^\dagger\} \) are the creation and annihilation operators associated with the \( j \)th nuclear DOF in the physical and tilde Hilbert spaces, respectively.

Substituting eqs 27 and 28 into eq 25, we obtain
\[
\frac{d}{dt} \psi_{0,\beta}(\beta, t) = -\frac{i}{\hbar} \Pi_\beta \psi_{0,\beta}(\beta, t)
\]
(29)
with \( \psi_{0,\beta}(\beta, 0) = |\gamma\rangle \otimes |0, 0\rangle \), \( \psi_{0,\beta}(\beta, t) = e^{i\beta t} \psi_{0,\beta}(\beta, t) \), and \( \Pi_\beta \) defined as
\[
\Pi_\beta = e^{i\beta} \Pi e^{-i\beta}
\]
(30)
The time-dependent thermal state \( \psi_{0,\beta}(\beta,t) \) is represented as
\[
\psi_{0,\beta}(\beta, t) = \sum_{l_1, \ldots, l_d} X(\beta, t; j_1, \ldots, j_d) |l_1 \otimes \ldots \otimes l_d\rangle
\]
(31)
where \( d \) is the overall number of DOF \((d = 1 + 2N_a)\) and \( \{j_\ell\} \) is the basis set with \( k = 1, \ldots, d \). We determined the size of the basis according to the convergence test, including two electronic state eigenvectors and the 10 nuclear harmonic eigenvectors for the nuclear DOF.

The time- and temperature-dependent expansion coefficients \( \{X(\beta, t; j_1, \ldots, j_d)\} \) correspond to an \( n_1 \times \cdots \times n_d \) complex array which requires storage space and computational effort that grows exponentially with \( d \). Thus, we avoid the curse of dimensionality by implementing the TFD wavepacket in the tensor-train (TT) format.

4.3. TT Format. The TT format of \( X \in \mathbb{C}^{n_1 \times \cdots \times n_d} \) involves a train-like product of \( d \) tensors, \( X_\ell \in \mathbb{C}^{n_\ell \times n_\ell} \), with \( r_0 = r_d = 1 \). Any particular element \( X(j_1, \ldots, j_d) \) can be evaluated by multiplication of the cores, as follows:
\[
X(j_1, \ldots, j_d) = \sum_{a_1=1}^{n_1} \sum_{a_2=1}^{n_2} \ldots \sum_{a_d=1}^{n_d} X_1(a_1, j_1, a_2) X_2(a_1, j_2, a_3) \ldots X_d(a_d, j_d)
\]
(32)
This can also be written in compact matrix product notation, as follows:
\[
X(j_1, \ldots, j_d) = X_1(j_1) X_2(j_2) \cdots X_d(j_d)
\]
(33)
with matrix \( X_\ell(j_\ell) \in \mathbb{C}^{n_\ell \times n_\ell} \) defining the \( j_\ell \)th slice of \( X_\ell \).

The central idea of the TT format is to generalize the concept of factorization. Each physical dimension \( i \) is factorized as an individual core (i.e., \( X_\ell \)). Entanglement with other physical dimensions is established through the auxiliary indices \( a_{i-1} \) and \( a_i \). The TT-ranks \( r_0, \ldots, r_d \) introduced by eq 32 remain small for a low level of entanglement and when they are \( r_0 = \cdots = r_d = 1 \), the TT format of \( X \) is a factorizable product.

Equation 32 shows that the TT format allows for compressed representations of \( X_\ell \), since it requires storage of \( X_\ell \) with \( n_\ell^2 \) elements when \( r_\ell = 1 \). Small \( F \) such a representation bypasses the need to explicitly store all \( n_\ell \) elements of \( X_\ell \) thus offering an exponential advantage in storage and computational effort.

In TT-TFD, the initial state \( \psi_{0,\beta}(\beta, 0) = |\gamma\rangle \otimes |0, 0\rangle \) takes an initial single-product form and is prepared as a rank-1 tensor train. The transformed TFD Schrödinger equation is then solved with the TT-KSL method.\(^{67,68}\) The TT-KSL propagator evolves the wavepacket according to the time-dependent variational principle (TDVP) by evolving the time-dependent state on a fixed-rank TT manifold. Comparisons to other TT propagators have shown that TT-KSL is quite accurate and efficient.\(^{53,69}\)

4.4. Projection-Free Inputs from TT-TFD. The PFIs required for calculating the memory kernel and inhomogeneous term of the GQME are computed by using the TT-TFD methodology. According to eq 16, the matrix elements \( \mathbf{U}_\ell(\tau) \) obtained as follows:
\[
\mathbf{U}_\ell(\tau) = \text{Tr}_{e}(e^{-\beta_1 t/\hbar} \rho_0(0) |m\rangle \langle m| e^{i\beta_1 t/\hbar} (\hbar/\sqrt{2})(\ell \otimes \hat{I}_e))
\]
(34)
Since TT-TFD requires an initial electronic state that is in a pure state \( |\gamma\rangle \), in the following, we write \( |m\rangle \langle m| \) as \( |\gamma\rangle \langle \gamma| \); however, we note that all \( \mathbf{U}_\ell(\tau) \) elements with off-diagonal initial electronic density matrices can be expressed as linear combinations of pure-state populations (see the SI).

We use \( \langle l| \langle j| (l \otimes \hat{I}_e) \rangle \) to rewrite \( \mathbf{U}_\ell(\tau) \) as
\[
\mathbf{U}_\ell(\tau) = \text{Tr}_{e}(e^{-\beta_1 t/\hbar} \rho_0(0) \gamma_l \gamma_j e^{i\beta_1 t/\hbar}(\ell \otimes \hat{I}_e))
\]
(35)
From this equation, noting that \( \text{Tr}_e[|\psi_\ell(\beta, \tau)\rangle \langle \psi_\ell(\beta, \tau)|] = e^{-\beta_1 t/\hbar} \rho_0(0) |\gamma| e^{i\beta_1 t/\hbar} \), we perform a cyclic permutation to obtain
\[
\mathbf{U}_\ell(\tau) = \text{Tr}_{e}(\text{Tr}_l(l \otimes \hat{I}_e) \psi_\ell(\beta, t) \psi_\ell(\beta, \tau)) (l \otimes \hat{I}_e))
\]
(36)
From here, we use \( \psi_\ell(\beta, \tau) = (l \otimes \hat{I}_e) \psi_\ell(\beta, \tau) \) and \( \psi_\ell(\beta, \tau) = e^{i\beta t} \psi_\ell(\beta, \tau) \) to obtain
\[
\mathbf{U}_\ell(\tau) = \text{Tr}_{e}(\text{Tr}_l(l \otimes \hat{I}_e) \psi_\ell(\beta, \tau)) (\ell \otimes \hat{I}_e))
\]
(37)
which provides the elements of \( \mathbf{U}(\tau) \) after obtaining \( \psi_\ell(\beta, \tau) \) by integrating eq 29.

5. APPLICATIONS
In this section, we report simulations of electronic population dynamics based on four types of GQMEs. The GQMEs correspond to different subsets of electronic reduced density matrix elements used to describe the underlying dynamics (see Section 3). As described in Section 4, the memory kernels and inhomogeneous terms are calculated from PFIs obtained via the quantum-mechanically exact TT-TFD method as applied to five different realizations of a benchmark spin-boson model Hamiltonian. We also compare the quantum-mechanically exact memory kernels and inhomogeneous terms obtained with TT-TFD inputs to calculations based on an approximate linearized semiclassical (LSC) method.\(^{19}\)

The reduced electronic density matrix for the spin-boson model, introduced in Section 5.1, consists of four matrix elements: \( \{\sigma_{DD}, \sigma_{DD}, \sigma_{AD}, \sigma_{AA}\} \) where |D) and |A) correspond to the donor and acceptor electronic states, respectively. We
consider GQMEs for the following four subsets of matrix elements: (1) \(\{\sigma_{DD}, \sigma_{DA}, \sigma_{AD}, \sigma_{AA}\}\) (the full density matrix); (2) \(\{\sigma_{DD}, \sigma_{AA}\}\) (the populations-only subset); (3) \(\{\sigma_{DD}\}\) (the donor single-population subset); and (4) \(\{\sigma_{AA}\}\) (the acceptor single-population subset). The TT-TFD-based PFIs, obtained by taking numerical derivatives of the time evolution operator \(U(t)\) [see eq 16], are compared to PFIs obtained via an LSCII-based method denoted LSCII [also referred to as the LSC initial value representation (LSC-IVR) method\(^\text{39}\)]. Reference 44 provides a detailed discussion of the protocols used for calculating PFIs via LSCII.

### 5.1. Spin-Boson Models

The spin-boson model provides a useful framework for studying molecular systems where the dynamics involves two coupled electronic states. In the simplest form, the electronic coupling is independent of the nuclear coordinates (the so-called “Condon approximation”). The nuclear motion in each electronic state is described by harmonic potential energy surfaces (PESs) with distinct equilibrium energies and equilibrium positions. As such, the spin-boson model has been widely used for describing a wide range of chemical dynamical processes, including charge and energy transfer (e.g., Marcus theory), nonadiabatic dynamics, photochemistry, spin energy relaxation and dephasing, vibrational energy relaxation, and, more recently, polaritonic chemistry where the photonic DOF can be described as harmonic oscillators and therefore grouped with the nuclear DOF.\(^\text{13-16,71}\)

The spin-boson Hamiltonian is defined according to eq 1 with \(\{\hat{H}_1\}\) and \(\{\hat{V}_{k} \rightarrow \hat{V}_{k}\}\), which are defined as follows:

\[
\hat{H}_1 \equiv \hat{H}_D = \epsilon + \sum_{k=1}^{N} \frac{\hat{P}_k^2}{2} + \frac{1}{2} \omega_k \hat{\hat{R}}_k^2 - c_k \hat{\hat{R}}_k^2
\]

\[
\hat{H}_2 \equiv \hat{H}_A = -\epsilon + \sum_{k=1}^{N} \frac{\hat{P}_k^2}{2} + \frac{1}{2} \omega_k \hat{\hat{R}}_k^2 + c_k \hat{\hat{R}}_k^2
\]

\[
\hat{V}_{12} \equiv \hat{V}_{DA} = \hat{V}_{21} \equiv \hat{V}_{AD} = \Gamma
\]

Here, \(2\epsilon\) is the energy difference between the donor (D) and acceptor (A) states with nuclear coordinates at equilibrium, and the electronic coupling between donor and acceptor states is defined by the positive constant \(\Gamma\) (the Condon approximation).

The discrete set of \(N_a\) frequencies \(\{\omega_k\}\) and electron–phonon coupling coefficients \(\{c_k\}\) of the nuclear modes are sampled from an Ohmic spectral density with an exponential cutoff:

\[
j(\omega) = \frac{\pi}{2} \sum_{k=1}^{N} \frac{c_k^2}{\omega_k} \delta(\omega - \omega_k) \xrightarrow{N_a \to \infty} \frac{\pi \hbar}{2} \xi \omega e^{-\omega/\omega_0}
\]

### 5.2. GQMEs

The following subsections outline four types of GQMEs examined by our simulations, corresponding to the analysis of quantum dynamics for different subsets of electronic reduced density matrix elements.

#### 5.2.1. Full GQME for All Electronic Density Matrix Elements

Here, we consider the GQME where the quantity...
of interest includes all four reduced electronic density matrix elements, \( \{ \sigma_{ab}(t) \} = \{ \sigma_{DD}(t), \sigma_{DA}(t), \sigma_{AD}(t), \sigma_{AA}(t) \} \):

\[
\frac{d}{dt} \sigma_{jk}(t) = -i \sum_{l,m=1}^{N=2} \langle L_{jk,lm} \rangle_{0} \sigma_{lm}(t) - \sum_{l,m=1}^{N=2} \int_{0}^{t} dt' \mathcal{K}^{\text{full}}_{jk,lm}(t) \sigma_{lm}(t - \tau)
\]

(43)

where \( jk \in \{ DD, DA, AD, AA \} \). The memory kernel superoperator \( \mathcal{K}^{\text{full}}(\tau) \) is represented by an \( N^2 \times N^2 = 4 \times 4 \) time-dependent matrix whose matrix elements are obtained by solving the following Volterra equation:

\[
\mathcal{K}^{\text{full}}_{jk,lm}(\tau) = i \mathcal{F}_{jk,lm}(\tau) - \frac{1}{\hbar} \sum_{u,v=1}^{N=2} \mathcal{F}_{jk,uv}(\tau) \langle L_{uv,lm} \rangle_{0} + i \sum_{u,v=1}^{N=2} \int_{0}^{t} dt' \mathcal{F}_{jk,uv}(\tau - \tau') \mathcal{K}^{\text{full}}_{uv,lm}(t')
\]

(44)

where the PFIs \( \{ \mathcal{F}_{jk,lm}(\tau) \} \) and \( \{ \mathcal{F}_{jk,lm}(\tau) \} \) are introduced by eq 12.

5.2.2. Populations-Only GQME for Diagonal Elements of the Reduced Electronic Density Matrix. Here, we consider the GQME for the quantity of interest includes only the diagonal matrix elements of the reduced electronic density matrix (i.e., the populations-only GQME), such that \( \{ \sigma_{ab}(t) \} = \{ \sigma_{DD}(t), \sigma_{AA}(t) \} \):

\[
\frac{d}{dt} \sigma_{jk}(t) = -i \sum_{l=1}^{N=2} \int_{0}^{t} dt' \mathcal{K}^{\text{pop}}_{jk,kl}(t) \sigma_{kl}(t - \tau)
\]

(45)

where \( j \in \{ D,A \} \). The memory kernel superoperator \( \mathcal{K}^{\text{pop}}(\tau) \) is represented by an \( N \times N = 2 \times 2 \) time-dependent matrix, with individual matrix elements obtained by solving the following Volterra equation:

\[
\mathcal{K}^{\text{pop}}_{jk,kl}(\tau) = i \mathcal{F}_{jk,kl}(\tau) + i \sum_{j,k=1}^{N=2} \int_{0}^{t} dt' \mathcal{F}_{jk,lk}(\tau - \tau') \mathcal{K}^{\text{pop}}_{jk,kl}(\tau')
\]

(46)

where the PFIs \( \{ \mathcal{F}_{jk,kl}(\tau) \} \) and \( \{ \mathcal{F}_{jk,kl}(\tau) \} \) are introduced by eq 12.

5.2.3. Single-Population Scalar GQMEs for One Diagonal Element of the Reduced Electronic Density Matrix. Finally, we consider the two single-population scalar GQMEs for the case where the subset includes either only the population of the donor state \( \{ \sigma_{DD}(t) \} \) or only the population of the acceptor state \( \{ \sigma_{AA}(t) \} \), such that \( \{ \sigma_{ab}(t) \} = \{ \sigma_{DD}(t) \} \) or \( \{ \sigma_{ab}(t) \} = \{ \sigma_{AA}(t) \} \), respectively:

\[
\frac{d}{dt} \sigma_{DD}(t) = -i \int_{0}^{t} dt' \mathcal{K}^{\text{donor}}_{DD,DD}(t) \sigma_{DD}(t - \tau)
\]

(47)

\[
\frac{d}{dt} \sigma_{AA}(t) = -i \int_{0}^{t} dt' \mathcal{K}^{\text{acceptor}}_{AA,AA}(t) \sigma_{AA}(t - \tau) + I_{AA}^{\text{acceptor}}(t)
\]

(48)

Note that the inhomogeneous term does not vanish in the case where \( \{ \sigma_{ab}(t) \} = \{ \sigma_{AA}(t) \} \). Also note that the memory kernels \( \mathcal{K}^{\text{donor}}_{DD,DD}(\tau) \) and \( \mathcal{K}^{\text{acceptor}}_{AA,AA}(\tau) \), as well the inhomogeneous term \( I_{AA}^{\text{acceptor}}(t) \), are scalar in this case and can be obtained by solving the following Volterra equations:

\[
\mathcal{K}^{\text{donor}}_{DD,DD}(\tau) = i \mathcal{F}^{\text{donor}}_{DD,DD}(\tau) + i \int_{0}^{t} dt' \mathcal{F}^{\text{donor}}_{DD,DD}(\tau - \tau') \mathcal{K}^{\text{donor}}_{DD,DD}(\tau')
\]

(49)

\[
\mathcal{K}^{\text{acceptor}}_{AA,AA}(\tau) = i \mathcal{F}^{\text{acceptor}}_{AA,AA}(\tau) + i \int_{0}^{t} dt' \mathcal{F}^{\text{acceptor}}_{AA,AA}(\tau - \tau') \mathcal{K}^{\text{acceptor}}_{AA,AA}(\tau')
\]

(50)

\[
I_{AA}^{\text{acceptor}}(t) = -i \mathcal{F}^{\text{acceptor}}_{AA,DD}(t) + i \int_{0}^{t} dt' \mathcal{F}^{\text{acceptor}}_{AA,DD}(t - \tau') \mathcal{K}^{\text{acceptor}}_{AA,DD}(t')
\]

(51)

where the PFIs \( \mathcal{F}^{\text{donor}}_{DD,DD}, \mathcal{F}^{\text{donor}}_{AA,AA}, \mathcal{F}^{\text{acceptor}}_{DD,DD}, \mathcal{F}^{\text{acceptor}}_{AA,AA}, \mathcal{F}^{\text{acceptor}}_{AA,DD}, \) and \( \mathcal{F}^{\text{acceptor}}_{AA,DD} \) are defined by eq 12.

5.3. Input Methods. It is important to note that the four types of GQMEs, outlined in the previous subsections, call for the same input of PFIs defined by eq 12. The different types of GQMEs differ only with respect to the specific matrix elements of \( \mathcal{F}(\tau) \) and \( \mathcal{F}(\tau) \) that are required to calculate the memory kernel and inhomogeneous term. For example, calculating the memory kernel for evolving the full set of reduced density matrix elements according to eq 43 requires calculation of all 16 matrix elements of \( \mathcal{F}(\tau) \) and \( \mathcal{F}(\tau) \). In contrast, calculating the memory kernel of the donor single-population GQME (eq 47) requires only a single matrix element of each of the matrices representing \( \mathcal{F}(\tau) \) and \( \mathcal{F}(\tau) \).

The matrix elements of \( \mathcal{F}(\tau) \) and \( \mathcal{F}(\tau) \) can be determined using a wide range of numerically exact or approximate propagation methods. Since the matrix elements of \( \mathcal{F}(\tau) \) and \( \mathcal{F}(\tau) \) are given in terms of two-time correlation functions of the overall system, the only requirement for a propagation method is that it should be able to calculate such quantities, either exactly or approximately.

In this paper, we compare and contrast two input methods: the quantum-mechanically exact TT-TFD method described in Section 4 and the approximate semiclassical LSCII method, previously described in ref 19. The inclusion of the LSCII input method is done for the sake of comparison between the memory kernels and inhomogeneous terms as obtained from an approximate input method with those obtained via an exact input method, with the intent of exploring the main sources of inaccuracy when approximate input methods are used.

For the LSCII method, we calculate \( \mathcal{F}^{\text{LSCII}}_{jk,lm}(\tau) \) and \( \mathcal{F}^{\text{LSCII}}_{jk,lm}(\tau) \) directly as described in ref 19. For the TT-TFD method, we calculate the \( N^2 \times N^2 \) elements of the time evolution operator of the electronic reduced density matrix \( \mathcal{U}(\tau) \) introduced by eq 16. Then, \( \mathcal{F}^{\text{TT-TFD}}_{jk,lm}(\tau) \) and \( \mathcal{F}^{\text{TT-TFD}}_{jk,lm}(\tau) \) are obtained from numerical derivatives according to eq 17. For the results given in this paper, the numerical derivatives were calculated using the second-order finite central difference method available in the NumPy Python library. The same time step is used for both LSCII and TT-TFD (see Table 1) and \( 10^6 \) trajectories were used for LSCII.

Once the PFIs have been obtained with either TT-TFD or LSCII propagation, the memory kernels and inhomogeneous terms of the GQMEs are calculated via an iterative algorithm that solves the corresponding Volterra equation (see eqs 44, 46, 49, 50, and 51). The different types of GQMEs (see eqs 43, 45, 47, and 48) are then solved numerically for the
5.4. Results. Figure 1 compares the time-dependent $\sigma_i(t) = \sigma_{DD}(t) - \sigma_{AA}(t)$, showing the differences of electronic populations for the five realizations of the spin-boson model outlined in Section 5.1 (see Table 1). These results are obtained by using the four different types of GQMEs outlined in Section 5.2, with PFIs computed with the TT-TFD method as described in Section 4. These results provide a clear demonstration of the rather remarkable fact that all four GQMEs correspond to exact equations of motion for the electronic populations and thereby reproduce the same exact population dynamics when a quantum-mechanically exact input method like TT-TFD is used, even though they are quite different in form and dimensionality.

Next, we focus on model 4 for a more-detailed analysis, with the analogous analysis for the other models provided in the SI. Figure 2 compares the population relaxation dynamics for model 4 (see Table 1), obtained with different types of GQMEs and memory kernels calculated by TT-TFD and LSCII input methods. The population relaxation dynamics generated via the LSCII-based populations-only GQME is in excellent agreement with the exact results. At the same time, the population relaxation dynamics generated via the TT-TFD method satisfy the Condon approximation, as derived in ref 43: 

$$\mathcal{K}_{ab,c,d}(\tau) = \frac{1}{\hbar} \text{Tr} \{ | b \rangle \langle a | \mathcal{L}_{\text{zero}} e^{-(Q/L)^{1/2} Q_L} \mathcal{L}_{\text{zero}}^d (0) | c \rangle \langle d | \}$$

(52)

where $\mathcal{L}_{\text{zero}}(\cdot) = \left[ \sum_{j=1}^N \hat{H}_j \right] (| j \rangle \langle j |) = [\hat{H}_{\text{zero}}, \cdot]$. For any system satisfying the Condon approximation, $\hat{V}_\text{gb} \rightarrow V_\text{gb}$ (such as the spin-boson model under consideration in this paper), eq 52 is equivalent to eq 44, with the proof given in Appendix A of ref 43. The fact that $| a \rangle \langle a | \mathcal{L}_{\text{zero}} = \left[ | a \rangle \langle a |, \sum_{j=1}^N \hat{H}_j \right] | j \rangle \langle j | = \hat{a}_a | a \rangle \langle a | \hat{a}_a^\dagger - \delta_{a,0} \hat{a}_a | a \rangle = 0$.
Importantly, while the DDAD, DDDA, AADA, and AAAD matrix elements in the first and fourth rows of Figure 2 reveal several trends:

- The agreement between TT-TFD and LSCII is significantly better for the matrix elements $\mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{DDAA}}^{\text{full}}, \mathcal{K}_{\text{ADDA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$ than for the matrix elements $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$. At the same time, the four matrix elements $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$ are significantly smaller than the remaining four matrix elements $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$. Thus, LSCII appears to capture the larger-amplitude matrix elements better than the smaller ones. Given the expectation that the larger-amplitude matrix elements would play a more significant role in the dynamics, this observation is consistent with the relative accuracy of the LSCII-based GQME.

- Whereas the real parts of the larger matrix elements $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$ are seen to be relatively short-lived (compared to the population relaxation time scale, see Figure 1) and exhibit a monotonic decay, the imaginary parts are seen to be oscillatory and do not appear to decay. Note that the oscillatory behavior of the imaginary parts obtained via LSCII is damped, compared to exact results obtained via TT-TFD. The observed damping is likely a manifestation of the quasiclassical nature of LSCII, which limits its ability to accurately capture coherent quantum dynamics. Since one expects the real parts to dominate population relaxation rates, the relative accuracy of the LSCII-based GQME can be attributed to the ability of LSCII to capture the real parts rather well.

Examination of the remaining nonvanishing matrix elements, $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$, reveals the following trends:

- The real parts of $\mathcal{K}_{\text{DDDA}}^{\text{full}}$ and $\mathcal{K}_{\text{DDAD}}^{\text{full}}$ are significantly larger and less oscillatory than the real parts of $\mathcal{K}_{\text{DDDA}}^{\text{full}}$ and $\mathcal{K}_{\text{DDAD}}^{\text{full}}$. This implies that the dynamics of the coherences $\sigma_{\text{DA}}$ and $\sigma_{\text{AD}}$ is dominated by dephasing (with rates dictated by $\mathcal{K}_{\text{DDDA}}^{\text{full}}$ and $\mathcal{K}_{\text{DDAD}}^{\text{full}}$) and that coherence-to-coherence transfer (with rates dictated by $\mathcal{K}_{\text{DDDA}}^{\text{full}}$ and $\mathcal{K}_{\text{DDAD}}^{\text{full}}$) is significantly slower than dephasing. This is consistent with the secular approximation (also called the rotating wave approximation), which is often invoked to eliminate coherence transfer terms from perturbative quantum master equations.73 LSCII appears to capture the real parts of $\mathcal{K}_{\text{DDDA}}^{\text{full}}, \mathcal{K}_{\text{DDAD}}^{\text{full}}, \mathcal{K}_{\text{AADA}}^{\text{full}}, \mathcal{K}_{\text{ADAD}}^{\text{full}}$ rather accurately. LSCII also appears to be less accurate when it comes to capturing the corresponding imaginary parts, with the inaccuracy manifested by an overdamping of the oscillatory behavior. This behavior is similar to that of...
noted above, regarding other matrix elements, and is consistent with the quasiclassical nature of the approximations on which LSCII is based.

Figure 3. Real parts of the matrix elements of the memory kernel of the GQME for the full electronic density matrix $[\mathbf{K}^{\rm full}(\tau)$ in eq 44] for model 4, as obtained from TT-TFD-based PFIs (solid blue lines) and LSCII-based PFIs (dashed red lines). Similar graphs for the other four models are provided in the SI.

Figure 4. Imaginary parts of the matrix elements of the memory kernel of the GQME for the full electronic density matrix $[\mathbf{K}^{\rm full}(\tau)$ in eq 44] for model 4, as obtained from TT-TFD-based PFIs (solid blue lines) and LSCII-based PFIs (dashed red lines). Similar graphs for the other four models are provided in the SI.

Given that population transfer is mediated by the coherences in the case of the full density matrix GQME, the accuracy of the real parts of the LSCII-based $\{\mathbf{K}_{\rm DADA}^{\rm full}, \mathbf{K}_{\rm DAAD}^{\rm full}, \mathbf{K}_{\rm ADDA}^{\rm full}, \mathbf{K}_{\rm ADAD}^{\rm full}\}$ likely plays an important
role in the ability of the LSCII-based GQME to accurately predict the population relaxation dynamics (see Figure 2).

Next, we consider the memory kernel in the case of the GQME for the electronic populations, \( K^{\text{pop}}(\tau) \) (see eqs 45 and 46). In this case, the memory kernel is given in terms of a 2 \( \times \) 2 matrix that consists of the memory kernel elements in Figures 5 and 6: \( \{ K_{\text{DD},\text{DD}}^{\text{pop}}, K_{\text{DD},\text{AA}}^{\text{pop}}, K_{\text{AA},\text{DD}}^{\text{pop}}, K_{\text{AA},\text{AA}}^{\text{pop}} \} \). The dimensionality of \( K^{\text{pop}}(\tau) \) should be contrasted with the \( K^{\text{full}}(\tau) \), for which the same four matrix elements vanish. Since the coherences have been projected out for this GQME, the memory kernel gives rise to direct coupling between populations, as opposed to population transfer being mediated by the coherences. As a result, donor-to-acceptor population transfer corresponds to a one-step process.

Comparison of the TT-TFD-based and LSCII-based real and imaginary parts of \( \{ K_{\text{DD},\text{DD}}^{\text{pop}}, K_{\text{DD},\text{AA}}^{\text{pop}}, K_{\text{AA},\text{DD}}^{\text{pop}}, K_{\text{AA},\text{AA}}^{\text{pop}} \} \) reveals the following notable trends:

- The real parts of those four memory kernel matrix elements are comparable in size and exhibit a damped oscillatory behavior that is longer-lived than the nonvanishing matrix elements of \( K^{\text{pop}}(\tau) \). This behavior is consistent with previous studies and can be traced back to the fact that, in this case, the memory kernel also must account for the impact of the projected-out electronic coherences on the electronic populations.
- LSCII is highly accurate when it comes to reproducing the real parts of the exact TT-TFD-based \( \{ K_{\text{DD},\text{DD}}^{\text{pop}}, K_{\text{DD},\text{AA}}^{\text{pop}}, K_{\text{AA},\text{DD}}^{\text{pop}}, K_{\text{AA},\text{AA}}^{\text{pop}} \} \). Given that the real parts of the \( K^{\text{pop}}(\tau) \) matrix elements dominate the population transfer kinetics, this observation is consistent with the previously made observation that the LSCII-based populations-only GQME can reproduce the population relaxation rather well.
- Whereas the imaginary parts of \( \{ K_{\text{DD},\text{DD}}^{\text{pop}}, K_{\text{DD},\text{AA}}^{\text{pop}}, K_{\text{AA},\text{DD}}^{\text{pop}}, K_{\text{AA},\text{AA}}^{\text{pop}} \} \) computed with TT-TFD vanish, the corresponding LSCII values do not. The discrepancy is due to errors in the calculation of \( \hat{F}_{ij,\text{nm}}(\tau) \) elements with LSCII, which generates a small real part for the \( \hat{F}_{ij,\text{nm}}(\tau) \) elements, which should be purely imaginary. However, the failure of LSCII to accurately predict the imaginary parts does not appear to impact the accuracy of the population transfer kinetics, since the imaginary parts are 2 orders of magnitude smaller than the real parts.

Finally, we consider the scalar memory kernels in the donor and acceptor single-population GQMEs, \( K_{\text{DD},\text{DD}}^{\text{ donor}}(\tau) \) and

![Figure 5. Real parts of the matrix elements of the memory kernels for the populations-only and single-population GQMEs for model 4, as obtained from TT-TFD-based PFIs and LSCII-based PFIs. Shown are the matrix elements of three different memory kernels: (1) The memory kernel of the populations-only GQME \( K^{\text{pop}}(\tau) \) in eq 46], which has four elements (DDDD, DDAA, AADD, and AAAA) and is depicted with solid cyan lines for the results from TT-TFD-based PFIs and dashed magenta lines for the results from LSCII-based PFIs; (2) and (3) The single-element memory kernels of the scalar single-population GQMEs \( K_{\text{DD},\text{DD}}^{\text{ donor}}(\tau) \) and \( K_{\text{AA},\text{AA}}^{\text{ acceptor}}(\tau) \), in eqs 49 and 50, respectively\], which are depicted in the DDDD and AAAA panels, respectively, with solid green lines for the results from TT-TFD-based PFIs and dashed yellow lines for the results from LSCII-based PFIs. Graphs with the results for the other four models are provided in the SI.


In this case, $K_{\text{DDD}}^{\text{donor}}(\tau)$ is given by the top-left corner element and $K_{\text{AAA}}^{\text{acceptor}}(\tau)$ is given by the bottom-right corner element in Figures 5 and 6. Comparison of the real and imaginary parts of $K_{\text{DDD}}^{\text{donor}}(\tau)$ and $K_{\text{AAA}}^{\text{acceptor}}(\tau)$ computed with TT-TFD and LSCII reveals the following notable trends:

- The real parts of $K_{\text{DDD}}^{\text{donor}}(\tau)$ and $K_{\text{AAA}}^{\text{acceptor}}(\tau)$ are comparable in size and exhibit a damped oscillatory behavior with a lifetime similar to that of the populations-only memory kernel elements.

- LSCII is highly accurate for reproducing the real part of the exact TT-TFD-based $K_{\text{DDD}}^{\text{donor}}(\tau)$. The accuracy is somewhat lower for reproducing the real part of $K_{\text{AAA}}^{\text{acceptor}}(\tau)$.

- While the imaginary parts of $K_{\text{DDD}}^{\text{donor}}(\tau)$ and $K_{\text{AAA}}^{\text{acceptor}}(\tau)$ computed with TT-TFD vanish, the corresponding LSCII values do not. However, the failure of LSCII to accurately predict the imaginary parts does not appear to impact the accuracy of the population transfer kinetics, since the imaginary parts are 2 orders of magnitude smaller than the real parts.

In Figure 7, we show the real part of the inhomogeneous term of the acceptor single-population GQME, $\tilde{I}_{\text{AAA}}(t)$, which is the only GQME with an inhomogeneous term considered in

![Figure 6](https://example.com/f6.png)  
**Figure 6.** Imaginary parts of the matrix elements of the memory kernels for the populations-only and single-population GQMEs for model 4, as obtained from TT-TFD-based PFIs and LSCII-based PFIs. Shown are the matrix elements of three different memory kernels: (1) The memory kernel of the populations-only GQME $[K_{\text{pop}}(\tau)]$ in eq 46, which has four elements (DDDD, DDAA, AADD, and AAAA) and is depicted with solid cyan lines for the results from TT-TFD-based PFIs and dashed magenta lines for the results from LSCII-based PFIs; (2 and 3) the single-element memory kernels of the scalar single-population GQMEs $[K_{\text{DD}}^{\text{donor}}(\tau)$ and $K_{\text{AA}}^{\text{acceptor}}(\tau)$, in eqs 49 and 50, respectively], which are depicted in the DDDD and AAAA panels, respectively, with solid green lines for the results from TT-TFD-based PFIs and dashed yellow lines for the results from LSCII-based PFIs. Graphs with the results for the other four models are provided in the SI.

![Figure 7](https://example.com/f7.png)  
**Figure 7.** Real part of $\tilde{I}_{\text{AA}}(\tau)$ (see eq S1) for model 4, as obtained from TT-TFD-based PFIs (solid blue lines) and LSCII-based PFIs (dashed red lines). Similar graphs for the other four models are included in the SI.
this paper. The imaginary component is not shown, because it is zero for the results from both TT-TFD-based and LSCII-based PFIs. In the figure, we see that the inhomogeneous term from LSCII-based PFIs is slightly overdamped, compared to the inhomogeneous term from the TT-TFD-based PFIs.

To understand the origin of the inaccuracies in the LSCII-based single-population GQMEs relative to the populations-only GQME, we note that any such inaccuracies must come from inaccuracies in $\mathcal{F}(\tau)$ and $\mathcal{F}(\tau)$, as the subsequent steps of the GQME approach are exact. To this end, in Figure 8, we show the imaginary components of the matrix elements of $\mathcal{F}(\tau)$, in Figures 9 and 10, the real and imaginary components of the matrix elements of $\mathcal{F}(\tau)$ that are used as PFIs to obtain the memory kernels for the single-population and populations-only GQMEs. The real parts of $\mathcal{F}(\tau)$ are not shown, because they are zero for these elements from both LSCII and TT-TFD. These figures clearly show that, although the LSCII-based $\mathcal{F}(\tau)$ and $\mathcal{F}(\tau)$ matrix elements can be rather accurate, there are significant deviations from the exact ones. The deviations are the origin of any inaccuracies in the memory kernels obtained from them.

We now show that, although the errors in $\mathcal{F}(\tau)$ affect the memory kernels of both single-population and populations-only GQMEs, the effect is weaker on the latter because of error cancellation. To see this difference in effect, we note that $\mathbf{K}^{\text{pop}}_{\text{DD,DD}}(\tau)$ and $\mathbf{K}^{\text{donor}}_{\text{DD,DD}}(\tau)$ are obtained from the PFIs via eq 46 and 49:

$$
\mathbf{K}^{\text{pop}}_{\text{DD,DD}}(\tau) = i\mathbf{F}_{\text{DD,DD}}(\tau) + i \int_0^\tau d\tau' \mathbf{F}_{\text{DD,DD}}(\tau - \tau')\mathbf{K}^{\text{pop}}_{\text{DD,DD}}(\tau') + \mathbf{F}_{\text{DD,DD}}(\tau - \tau')\mathbf{K}^{\text{pop}}_{\text{AA,DD}}(\tau') \tag{54}
$$

$$
\mathbf{K}^{\text{donor}}_{\text{DD,DD}}(\tau) = i\mathbf{F}_{\text{DD,DD}}(\tau) + i \int_0^\tau d\tau' \mathbf{F}_{\text{DD,DD}}(\tau - \tau')\mathbf{K}^{\text{donor}}_{\text{DD,DD}}(\tau') \tag{55}
$$

Importantly, the integrand on the right-hand side of eq 54 gives rise to inherent error cancellation, since $\mathbf{K}^{\text{pop}}_{\text{DD,DD}}(\tau)$ and $\mathbf{K}^{\text{pop}}_{\text{AA,DD}}(\tau)$ are of opposite sign, which causes the errors in $\mathcal{F}(\tau)$ to cancel when they are correlated. Such correlation is observed for all of the models that we studied, e.g., in Figure 8, most evidently at $t = 1.9 \Gamma$, the approximate method significantly overestimated both $\mathcal{F}_{\text{DD,DD}}(\tau)$ and $\mathcal{F}_{\text{DD,AA}}(\tau)$.

On the other hand, eq 55 does not allow for such error cancellation, thereby making the single-population GQMEs less accurate than the populations-only GQME.

5.5. Computational Cost. In this section, we examine the scaling of the computational cost of the GQME approach with TT-TFD as the input method, with respect to GQME type.

We begin by considering the time step used to calculate the TT-TFD-based PFIs to obtain the converged memory kernel and the inhomogeneous term. In contrast to LSCII, which required a similar time step for all GQMEs, the time step needed for convergence is found to decrease with decreasing dimensionality. More specifically, although the results shown above are all for a time step of $\Delta t = 0.00150083 \Gamma^{-1}$ (except for model 4, see Table 1), the time step needed for convergence for the full density matrix GQME is in the
range of $\Delta t = 0.00300166 - 0.00450249 \Gamma^{-1}$, in contrast to the time step of $\Delta t = 0.00150083 \Gamma^{-1}$ that was required in the case of the populations-only and single-population GQMEs (except for the populations-only GQME of model 4, which required a time step of $\Delta t = 0.000750415 \Gamma^{-1}$).

In ref 19, we noted that the direct calculation of $\langle j | k \rangle$ given in eq 12 required calculation of the dynamics for more electronic initial conditions than only $| j \rangle \langle k |$ due to terms involving off-diagonal components of the Hamiltonian in the initial state. However, although direct calculation of $\langle j | k, lm \rangle$ is necessary when using approximate input methods, when using exact input methods, we can obtain $\langle j | k, lm \rangle$ as described in eq 17. Therefore, we only need to calculate the dynamics for the initial electronic state $| j \rangle \langle k |$ to obtain $\langle j | k, lm \rangle$ and, subsequently, $\langle j | k, lm \rangle$ and $\langle j | k, lm \rangle$ through eq 17. As a result, there is a significant reduction in the number of initial electronic states necessary to calculate the PFIs needed for the reduced-dimensionality GQMEs compared to the full GQME. More specifically, although the full GQME approach requires simulating the dynamics for four initial electronic states in the case of a two-state system, the populations-only GQME requires only two initial electronic states, the acceptor single-population GQME approach requires two initial electronic states (with one of them due to the inhomogeneous term), and the donor single-population GQME requires only one initial electronic state. Thus, reduced-dimensionality GQMEs significantly enhance computational efficiency with regard to the number of initial states that must be simulated when exact input methods such as TT-TFD are used.

Next, we consider the cost of obtaining the memory kernels from the PFIs. The computational complexity of each iteration in the Volterra algorithm for the memory kernel is expected to be $O(N_{\text{mat}}^3)$, where $N_{\text{mat}}$ is the number of matrix elements in a row of the memory kernel matrix (e.g., $N_{\text{mat}} = N_e^2$ for the full GQME, $N_{\text{mat}} = N_e$ for the populations-only GQME, and $N_{\text{mat}} = 1$ for the single-population GQMEs). This is true regardless of the input method used and, therefore, the cost of each iteration of the Volterra algorithm increases dramatically with memory kernel size. The computational complexity of each iteration in the Volterra algorithm for the inhomogeneous term scales more favorably at $O(N_{\text{mat}}^2)$ but may still become restrictive with increasing dimensionality. However, note that the inhomogeneous term often is not needed for the larger-dimensional full and populations-only GQME approaches.

The number of iterations required for the iterative Volterra algorithm for the memory kernel to converge is also rather sensitive to the type of GQME and the dimensionality of the electronic observable of interest. More specifically, whereas 2 iterations are required for calculating the single-population memory kernels and 2–3 iterations are needed in the case of the populations-only memory kernel for all the models, 5–7 iterations are required for the full GQME approach. An inhomogeneous term is only required for the acceptor single-population GQME approach and would be required for any GQME approach where the set of electronic states that it projects onto does not include the initial electronic state. Because of the scaling of the Volterra algorithm for the
inhomogeneous term, it is generally only favorable to use a GQME approach that requires an inhomogeneous term when the dimensionality of the set of electronic states projected onto is small.

The converged memory time for each of the models and GQME types is found using the algorithm outlined in the SI of ref 19. The basic premise of the algorithm is to first calculate the dynamics at the highest possible memory time, $t_{\text{mem,max}}$, based on the maximum time of the PFI dynamics and then proceed backward in memory time to find the shortest memory time that keeps each element and time step of the electronic density matrix within a convergence parameter when compared to the same element and time step of the dynamics with the highest possible memory time. For the models studied in this paper, the highest possible memory time was $t_{\text{mem,max}} = 15 \Gamma^{-1}$. The converged memory time for each model and GQME approach is given in Table 2. In agreement with the results for LSCII in ref 19, the full GQME typically corresponds to the shortest memory time and the reduced-dimensionality GQMEs require significantly longer memory times, particularly the single-population GQMEs. Whereas the RK4 algorithm is expected to have computational complexity $O(t_{\text{mem}})$, the cost of a single iteration of the Volterra algorithm for the memory kernel has quadratic computational complexity $O(t_{\text{mem}}^2)$. Thus, situations where the reduced dimensionality of

![TT-TFD vs. LSCII for the Imag Part of $\tilde{F}_{ji, kk}(\tau)$ for Model 4](image_url)

Figure 10. Imaginary parts of the DDDD, DDAA, AADD, and AAAA matrix elements of $\tilde{F}(\tau)$ (see eq 12) for model 4, as obtained via TT-TFD (solid blue lines) and LSCII (dashed red lines). Similar graphs for the other four models are included in the SI.

Table 2. Memory Time of Each GQME Approach for Each Model

| Model # | Input Method | Full GQME | Populations-Only GQME | Donor GQME | Acceptor GQME |
|---------|--------------|-----------|------------------------|------------|--------------|
| 1       | TT-TFD       | 5.5034    | 14.7534                | 14.7534    | 14.5034      |
|         | LSCII        | 5.25415   | 14.7541                | 14.5041    | 14.5041      |
| 2       | TT-TFD       | 1.65348   | 14.4035                | 14.9035    | 14.4035      |
|         | LSCII        | 3.00415   | 14.2541                | 14.2541    | 14.2541      |
| 3       | TT-TFD       | 9.5034    | 13.7534                | 13.5034    | 14.0034      |
|         | LSCII        | 7.25415   | 9.25415                | 12.0041    | 11.5041      |
| 4       | TT-TFD       | 14.6539   | 5.65386                | 14.9039    | 11.9039      |
|         | LSCII        | 4.09415   | 5.75415                | 13.5041    | 8.75415      |
| 5       | TT-TFD       | 9.65348   | 10.4035                | 13.9035    | 13.6535      |
|         | LSCII        | 12.7541   | 14.7541                | 14.2541    | 13.5041      |

“In this table, the colors are to provide a visual aid, with red indicating a memory time above $12 \Gamma^{-1}$, yellow indicating a memory time from $9–12 \Gamma^{-1}$, and green indicating a memory time below $9\Gamma^{-1}$.
6. CONCLUDING REMARKS

We have implemented the TT-TFD method to obtain quantum-mechanically exact memory kernels and inhomogeneous terms for different types of GQMEs that describe the dynamics of electronic DOF for the spin-boson model. We have analyzed a GQME for the four-element full electronic dynamics of electronic DOF for the spin-boson model. We have also demonstrated that all four GQMEs are exact equations of motion and thus reproduce the same exact population dynamics when parameterized by a quantum-mechanically exact input method such as TT-TFD, although the four GQMEs are different in form and dimensionality.

Advancing the capability to calculate quantum-mechanically exact memory kernels and inhomogeneous terms for different types of GQMEs is highly desirable for multiple reasons:

- First, note that the memory kernels and inhomogeneous terms in the case of quantum open systems serve a similar role to that of the Hamiltonian in the case of closed quantum systems. More specifically, similar to how analyzing the properties of the Hamiltonian is often used to shed light on the closed quantum system dynamics it gives rise to, one expects that knowing what the quantum-mechanically exact memory kernel and inhomogeneous term look like and how they depend on various parameters and different choices of projections could shed light on the open quantum system dynamics they give rise to.

- Second, quantum-mechanically exact memory kernels and inhomogeneous terms are particularly valuable to evaluate the capabilities of PFIs obtained with approximate input methods, as shown in our comparisons of memory kernels and inhomogeneous terms obtained with exact TT-TFD and approximate LSCII simulation methods.

- Third, quantum-mechanically exact memory kernels can be used as benchmarks to assess the accuracy of various types of perturbative quantum master equations (QMEs). More specifically, while the GQMEs correspond to the exact equations of motion of the subset of dynamical quantities of interest, the derivation of perturbative QMEs relies on approximate expressions for the memory kernels that are based on treating various terms in the Hamiltonian, such as the system-bath coupling or electronic coupling, as small perturbations. Thus, comparisons of the perturbative memory kernels to the exact kernels can provide a better understanding of the accuracy of perturbative methods and their range of validity.

- Fourth, in certain situations, simulating the quantum dynamics via a GQME may be more cost-effective than the direct use of the numerically exact quantum dynamics method. More specifically, restricting the use of a quantum-mechanically exact method with regard to calculating the PFIs can provide a more efficient route to obtain the dynamics of the quantity of interest, compared to extracting it from the overall system dynamics. The computational cost analysis of the TT-TFD-based GQME approach provided in this paper constitutes an important step toward understanding when and how simulating the quantum dynamics via a GQME approach is advantageous, compared to the direct use of the numerically exact quantum dynamics method.

Various extensions of this study would be highly desirable, including combining the GQME approach with other quantum-mechanically exact and approximate input methods, calculating memory kernels and inhomogeneous terms for other types of dynamical quantities of interest, and exploring the capabilities of the GQMEs on other benchmark models. Work on such extensions is currently underway and will be reported in future publications.

ASSOCIATED CONTENT

Data Availability Statement
The code for the TT-TFD + GQME simulation of Model 1 is available at https://github.com/NingyiLyu/TT-TFD-GQME. The data that supports the findings of this study are available within the article and SI.

Supporting Information
The Supporting Information is available free of charge at https://pubs.acs.org/doi/10.1021/acs.jctc.2c00892.

Thorough derivation of the GQME reduced-dimensionality approach and a description of the linear combinations used within the TT-TFD approach to obtain elements of the time evolution operator of the reduced electronic density operator, \( \mathcal{U}(t) \), for off-diagonal initial states, along with graphs of the PFIs, memory kernels, and inhomogeneous terms (PDF)

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Notes

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