Inverting pump-probe spectroscopy for state tomography of excitonic systems

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We propose a two-step protocol for inverting ultrafast spectroscopy experiments on a molecular aggregate to extract the time-evolution of the excited state density matrix. The first step is a deconvolution of the experimental signal to determine a pump-dependent response function. The second step inverts this response function to obtain the quantum state of the system, given a model for how the system evolves following the probe interaction. We demonstrate this inversion analytically and numerically for a dimer model system, and evaluate the feasibility of scaling it to larger molecular aggregates such as photosynthetic protein-pigment complexes. Our scheme provides a direct alternative to the approach of determining all Hamiltonian parameters and then simulating excited state dynamics. © 2013 AIP Publishing LLC. [http://dx.doi.org/10.1063/1.4800800]

I. INTRODUCTION

Ultrafast nonlinear spectroscopy allows us to experimentally observe excited state dynamics in molecular aggregates, and in particular, energy transfer essential to the function of natural light harvesting systems.\textsuperscript{1–3} The existence of these experimental tools prompts a natural question: is it possible to use spectroscopic measurements to directly infer the excited state of such systems? A complete answer to this question would be a procedure for quantum state tomography (QST), that is, for reconstruction of the full density matrix describing the quantum state.\textsuperscript{4,5} State tomography is a technique that has found widespread application for validating and characterizing quantum devices designed as components for quantum computation. Such full characterization of an exciton state over multiple pigments, beyond a mere classical probability distribution, would offer information essential to understanding the explicitly quantum features of energy transport, which include coherence,\textsuperscript{6} entanglement,\textsuperscript{7} and possibly other types of non-trivial quantum dynamics.\textsuperscript{8–11} In this work, we show that under appropriate conditions and assumptions, QST of excited states can be performed from the results of a series of pump-probe type ultrafast spectroscopies.

The most sophisticated nonlinear technique for resolving energy transfer dynamics is the two-dimensional (2D) photon-echo, in which the time delays between three ultrafast pulses are manipulated to provide a 2D map between pump and probe frequencies at fixed time delays.\textsuperscript{12,13} These two-dimensional maps allow for direct visualization of the relationship between excitation and emission energies as a function of delay time. More formally, 2D spectroscopy is usually interpreted in the limit of impulsive interactions. In this approximation, it provides snapshots of the 3rd-order response function.\textsuperscript{1} Important applications of 2D spectroscopy to photosynthetic systems have included resolving energy transfer pathways\textsuperscript{14} and the dynamics of electronic quantum beats.\textsuperscript{6,15–17} In contrast, pump-probe spectroscopy (also known as transient absorption) is a simpler type of 3rd-order spectroscopy that historically predates 2D. In a pump-probe setup, a pump pulse excites the system, which is probed at some time later by a probe pulse. Because of its relative ease of experimental implementation, pump-probe was used to follow ultrafast energy transfer dynamics in photosynthesis long before 2D spectroscopy. For example, it provided the first evidence of electronic quantum beats in photosynthetic pigment-protein complexes, in 1997.\textsuperscript{18} Pump-probe provides less information than the 2D photon echo, because the pump-probe signal can be obtained by integrating over the excitation axis in a 2D spectra.\textsuperscript{19} However, for the purposes of this work, pump-probe has a clear advantage, namely, that it can be directly interpreted as a measurement of the state created by the pump pulse. Formally, the pump dependence is entirely contained within the change in the density matrix of the system after interacting with the pump.\textsuperscript{20}

In the past, time-resolved spectra such as pump-probe have been analyzed by simultaneously or concurrently fitting spectral components, known as decay associated or species associated spectra, with a kinetic model.\textsuperscript{21,22} These techniques are powerful, as evidenced by their widespread application to experiments. However, much of the kinetic information they reveal can be seen more directly in 2D spectra. Moreover, kinetic models, although adequate for many purposes, cannot describe more complex dynamics, such as those deriving from quantum beats or from a non-Markovian bath. Our approach to QST side-steps the issues of extending such integrated analyses by focusing on identifying the quantum state directly.

Recently, it was shown that a combination of photon-echo measurements of excitonic systems can be combined to

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perform quantum process tomography of excitonic dimers, either by using differently colored pulses or by combining peak amplitudes from a set of 2D spectra. Process tomography is more general than state tomography, because it specifies the full set of possible quantum evolutions for a system given any initial condition. This makes it well suited to characterizing gates for quantum computation, which are by definition designed to handle any possible input state. However, determining the full process matrix is expensive: it requires at least \( d^2 \) real parameters for a \( d \)-dimensional Hilbert space, in contrast to \( d^3 \) parameters for state tomography. Moreover, for analysis of complex molecular dynamics in condensed phases, such information, although potentially helpful, is not necessary, because most trajectories contained in the process matrix start from initial conditions that are implausible for a molecular aggregate. Indeed, typical theoretical investigations of dynamics in light harvesting systems follow dynamics after excitation for only a limited set of plausible initial states, such as the states which absorb sunlight or excitations from neighboring antenna complexes. Finally, the relative simplicity of state tomography helps to simplify consideration of new theoretical approaches to tomography, particularly because process tomography is often based on a series of state tomographies.

In this paper, we present a new approach to state tomography of excitonic systems based on pump-probe spectroscopy. Our approach is based on a two stage protocol that separates the easy (field based) and hard (system based) parts of the inversion process. This yields several advantages over prior approaches, including the ability to use arbitrarily shaped laser pulses and to perform the first inversion even when the second inversion is not possible. After presenting the details of each of these inversions, we demonstrate their feasibility by applying them to invert the simulated spectra of a model dimer with a Markovian environment. We close with a consideration of the conditions under which state inversion would be feasible for a natural light-harvesting system, the FMO complex of green sulfur bacteria.

II. RECIPE FOR PUMP-PROBE SPECTROSCOPY

We begin by presenting the specific theoretical formalism for pump-probe spectroscopy that we propose to invert. The measured signal in any 3rd-order spectroscopy experiment, including pump-probe, is a function of the 3rd-order polarization. This 3rd-order polarization depends on three interactions between the applied fields and the sample, with the time-ordering of these interactions enforced by time delays of the pulses and by looking at the signal emitted in a particular phase-matched direction. For a pump-probe experiment, the first two interactions happen with the same pulse, the pump, and the last interaction is with the probe pulse. The phase-matched condition is that the signal is observed in the direction of the probe. Based on this phase-matched geometry and the response function formalism (see Appendix A), we can combine the allowed time orderings to write the nonlinear polarization for a pump-probe experiment under the rotating wave approximation as

\[
P^{(3)}(t) = \int_0^\infty dt_3 R_{pp}(t_3, \rho_{pp}^{(2)}(t - t_3)) E_{pp}^+(t - t_3),
\]

in terms of the pump-probe response

\[
R_{pp}(t, \rho_{pp}^{(2)}) = \frac{i}{\hbar} \text{Tr} \left[ \mu^{(-)} G(t) V^{(+)}(t_3) \rho_{pp}^{(2)} \right].
\]

This pump-probe response is of identical form to the linear response function, but with the electronic ground-state density matrix \( \rho_0 \) replaced by the second-order contribution to the density matrix \( \rho_{pp}^{(2)} \) that contributes to the phase-matched signal observed in a pump-probe experiment (that is, with signal wave-vector \( k_\text{S} = k_\text{pr} \)). The quantities \( E_{pp}^+(t) \) and \( E_{pp}^-(t) \) denote the complex envelopes of the pump and probe pulses, respectively, with \( E^{-} = (E^{+})^* \). The dipole operators \( \mu^{(-)} = \sum_n d_n a_n \) and \( \mu^{(+)} = (\mu^{(-)})^\dagger \) with \( a_n \) as the annihilation operator for an electronic excitation on pigment \( n \) and \( d_n \) the corresponding dipole moments. The Liouville space operator \( G(t) \) is the retarded material Green function for evolution for time \( t \) and \( V^{(\pm)} = [\mu^{(\pm)}, \cdot] \). Formally, the portion of the second order contribution to the density matrix which contributes to the signal is given by

\[
\rho_{pp}^{(2)}(t) = \left( \frac{i}{\hbar} \right)^2 \sum_\pm \int_0^\infty dt_2 dt_1 G(t_2) V^{(\pm)} G(t_1) \rho_0 \times E_{pp}^{\pm}(t - t_2) E_{pp}^{\mp}(t - t_2 - t_1).
\]

In deriving Eqs. (1)–(3), we employed the rotating wave approximation (accurate for resonant excitation) and neglected those terms from the double-quantum-coherence contribution \( k_\text{S} \neq k_\text{pr} \). Accordingly, we can safely neglect the possibility of multiple excitations in the calculation of \( \rho_{pp}^{(2)} \). In Appendix B we prove that the excited state portion of \( \rho_{pp}^{(2)} \) is both equal to the excited state portion of the full density matrix and is itself a valid (but unnormalized) density matrix.

The core of our proposed quantum state tomography is the sequential inversion of Eqs. (1)–(3). The remainder of this section discusses additional details relevant to simulating experimental signals to test our inversion procedure. We emphasize that these expressions hold under the very general conditions, requiring only the rotating wave approximation, that all applied fields are weak and negligible overlap between pump and probe pulses. No assumptions were made concerning the shapes of these pump and probe pulses. Our decomposition here is similar to the window-doorway picture for the pump-probe signal, but here we have separated out the influence of the control fields, even when not in the impulsive “snapshot” limit. Related expressions in terms of a convolution of pump and probe components have been shown to facilitate analysis of pump-probe experiments with shaped probes.

A. Detection scheme and probe convolution

In a typical pump-probe experiment, the probe pulse has a fixed time-envelope, subject to a variable delay time \( T \) between the two pulses. Accordingly, we may substitute \( E_{pp}^+(t) = E_{pp}(t - T) \). Likewise, experimental signals are most directly interpreted in the frequency domain, so we now
consider the Fourier transform of the nonlinear polarization, $P^{(3)}(\omega) = \int dt \, e^{i\omega(t - T)}P^{(3)}(t)$, calculated relative to the probe delay $T$. We can also write the pump-probe response in the Fourier domain, $R_{PP}(\omega, \rho_{PP}^{(2)}) = \int dt \, e^{i\omega t} R_{PP}(t, \rho_{PP}^{(2)})$. In terms of these quantities in the frequency domain with the explicit probe delay $T$, we can then replace Eq. (1) with a one-dimensional convolution,

$$P^{(3)}(\omega, T) = \int_{-\infty}^{\infty} d\tau \, R_{PP}(\omega, \rho_{PP}^{(2)}(\tau))E_{pr}(\tau - T)e^{i\omega(\tau - T)}$$.

(4)

To obtain this relation, we substituted $t_1 = t - \tau$ and extended the lower limit of the integral in Eq. (1) to $-\infty$, because by definition $G(t) = 0$ for $t < 0$. We cannot simply turn this convolution into a multiplication by taking the Fourier transform of these quantities with respect to $T$, because for small or negative delay times $T$, there are contributions from signals where the pump does not necessarily interact before the probe. If $R_{PP}(\omega, \rho_{PP}^{(2)}(\tau))$ does not vary appreciably over the duration of the probe pulse, then the equation above is the Fourier transform of the probe field envelope, so we can approximate

$$P^{(3)}(\omega, T) \approx E_{pr}(\omega)R_{PP}(\omega, \rho_{PP}^{(2)}(T))$$.

(5)

In the limit of a completely impulsive probe, $E_{pr}(t) \approx E_{ob}(t)$ and thus $E_{pr}(\omega)$ is constant, so the nonlinear polarization and the pump-probe response are equal up to a constant of proportionality.

We measure the nonlinear polarization $P^{(3)}(t)$ by detecting the corresponding signal field $E_{S}(t) \propto iP^{(3)}(t)$. Here we consider heterodyne detection, either with the probe pulse as in a standard “self-heterodyned” pump-probe setup, or with a separate local-oscillator (LO) pulse. The use of a separate local-oscillator is possible in a transient-grating setup, in which the pump pulse is replaced by two otherwise identical pumps with different wavevectors $k_1$ and $k_2$, so that the signal wavevector $k_S = -k_1 + k_2 + k_3$ does not match the probe wavevector $k_3$. Mathematically, this transient-grating signal yields the same nonlinear polarization as in pump-probe, although it raises experimental complications by requiring phase-stability with an additional pulse. In heterodyne detection, the absolute value squared of the sum of the signal and local-oscillator (or probe) fields can be spectrally dispersed and measured in the frequency domain. Typically, the signal field is much smaller than that of the local oscillator, so upon subtracting away the local oscillator contribution, the measured signal $S(\omega)$ is proportional to $\text{Re}[E_S(\omega) E_{LO}^{*}(\omega)]$, and thus

$$S(\omega, T) \propto |E_{pr}(\omega)|^2 \text{Im}R_{PP}(\omega, \rho_{PP}^{(2)}(T))$$.

(7)

In this case, the signal only depends on the imaginary (absorptive) part of the nonlinear polarization $P^{(3)}$ and the pump-probe response. In the alternative transient grating setup, as long as one is still in the limit of a fast probe, applying a $\pi/2$ phase shift to the now distinct local oscillator pulse allows for obtaining the real (dispersive) part of the pump-probe response function in a similarly direct manner. More generally, heterodyne detection with and without a $\pi/2$ phase shift allows for obtaining both real and imaginary parts of the nonlinear polarization, respectively.

### B. Pump-probe response function

To isolate the effect of the probe, the pump-probe response function given by Eq. (2) can be written as

$$R_{PP}(t, \rho_{PP}^{(2)}) = \text{Tr} \left[ \hat{P}(t) \rho_{PP}^{(2)} \right]$$,

(8)

with the pump-probe response operator $\hat{P}(t)$ defined as

$$\hat{P}(t) = \frac{i}{\hbar} \mu^{(-)}(t) \hat{G}(t) V^{(+)}$$.

(9)

When inserted in Eq. (8), the action of $\hat{P}(t)$ is equivalent to the action of $i/\hbar [\mu^{(-)}(t), \mu^{(+)}(0)]$, where $\mu^{(\pm)}(t)$ denotes $\mu^{(\pm)}$ in the Heisenberg picture. This is similar but not equivalent to a family of quantum measurements parametrized by the continuous time variable $t$ (or frequency $\omega$ in the Fourier domain), since $R_{PP}$ can be complex valued. Accordingly, the pump-probe response can be interpreted as the projection of $\rho_{PP}^{(2)}$ onto $\hat{P}(t)$, where these are viewed as vectors in Liouville space,

$$R_{PP}(t, \rho_{PP}^{(2)}) = \langle \hat{P}(t) | \rho_{PP}^{(2)} \rangle$$.

(10)

Individual components $\langle \hat{P}(\omega)| \alpha \rangle$ of the pump-probe response operator are equivalent to the species associated spectra of the state $|\alpha\rangle$.

In most spectroscopy experiments, the signal is an ensemble measurement summed over all possible molecular orientations and static disorder of Hamiltonian parameters (inhomogeneous broadening). Accordingly, the pump-probe response in Eq. (2) should be replaced by its average over molecular orientations and static disorder. The orientational average can be handled elegantly using the expression for the pump-probe response in Eq. (8): in the magic angle $\theta \approx 54.7^\circ$ (MA) relative polarization configuration between the pump and probe pulses, the quantities $\hat{P}(t)$ and $\rho_{PP}^{(2)}$ can simply be replaced by their independent isotropic averages,

$$\langle R_{PP}(t, \rho_{PP}^{(2)}) \rangle_{\text{MA}} = \text{Tr} \left[ \langle \hat{P}(t) \rangle_{\text{iso}} \rho_{PP}^{(2)}_{\text{iso}} \right]$$.

(11)

By virtue of the properties of isotropically averaged tensors, these independent isotropic averages are equal to the average of the quantities obtained from the $xx$, $yy$, and $zz$ configurations. In contrast, the ensemble average over static disorder cannot be factorized this way in general, because under static disorder the pump-probe operator and density matrix are correlated, and altering the system Hamiltonian (e.g., to shift transition energies) changes both quantities systematically.

### III. INVERSION PROTOCOLS

#### A. Deconvolution of the pump-probe signal

The first stage of our inversion protocol is a double-deconvolution to determine the complex valued pump-probe
response function $R_{pp}(T, \rho_{pp}^{(2)})$ from the results of a series of heterodyne measurements, i.e., the signal $S(\omega, T)$. We need such a double-deconvolution procedure because the results of heterodyne detection depend on a trivial convolution over the nonlinear polarization, which in turn depends on a convolution over the response function [see Eqs. (4) and (6)]. Since the excited state density matrix is entirely contained in the pump-probe response function (Appendix B), this inversion retains all information about the quantum state. However, it is not immediately clear that the real (dispersive) part of the response function contains useful information independent of the imaginary (absorptive) part, which is the portion measured by usual pump-probe experiments.

Inverting the signal to obtain the pump-probe response function is a non-trivial but important task, since, as pointed out above, the signal is directly proportional to the response only when the probe pulse is much faster than all energy transfer dynamics. Such pulses can be difficult to realize experimentally. The need for a full inversion to obtain the response function is particularly relevant for understanding experiments which show fast oscillations due to quantum beats, whether these are of electronic, vibrational or mixed origin. In such cases, the fast probe assumption of Eq. (5) is not valid. We shall refer to the use of this approximate description for inversion as the “naïve” approach. In contrast, a proper treatment of this inverse problem would attempt to undo the convolution in Eq. (4).

To address this challenge, we suggest the use of standard deconvolution techniques based on general-form Tikhonov regularization (also known as ridge regression), which we describe in detail in Appendices C and D. The response function can then be obtained from two sequential 1D deconvolutions. First, we invert the measured signal $S(\omega, T)$ recorded at each choice of delay time $T$ to obtain the nonlinear polarization $P^{(3)}(\omega, T)$. The relationship between these signals is simple multiplication by the probe (or local oscillator) field in the frequency domain, so this step only uses the deconvolution to smooth the reconstruction along the $\omega$-axis. Second, we invert the nonlinear polarization $P(\omega, T)$ with a one-dimensional deconvolution for each fixed value of $\omega$ to obtain the response function $R_{pp}(\omega, \rho_{pp}^{(2)}(T))$. For this inversion, we only use experimental data with the delay between pump and probe pulses long enough so that we can ignore pulse overlap effects. Otherwise, we would be including non-pump-probe contributions to the signal. However, we nevertheless also reconstruct the pump-probe response at shorter times in order to appropriately handle boundary conditions, since the probe convolution means that these values for the response function contribute to the nonlinear polarization inside our region of interest.

This first stage in the inversion of pump-probe experiments requires only the detection results, i.e., the signal $S(\omega, T)$, and an excellent characterization of the probe and local oscillator fields. No system information is required at all. Likewise, we have sacrificed no information from our measurement about the internal system information, including its quantum state. Thus, in principle, this stage can be performed with high accuracy for any system, no matter how complex its internal degrees of freedom.

### B. Obtaining the quantum state

The second step to complete the state tomography is to invert the pump-probe response function $R_{pp}(\omega, \rho_{pp}^{(2)}(T))$ to obtain the quantum state $\rho_{eq}(T)$. This is certainly the harder step, since it requires the ability to construct the pump-probe response of arbitrary states. The necessary information is contained in the pump-probe operator $P(t)$ given by Eq. (9); calculating this requires both the transition dipole moments and a model for dynamics of the 1-exciton coherences between the probe and signal interactions. Essentially the same information is necessary to implement proposed algorithms for quantum process tomography. However, we emphasize that we do not need to know the nature of the initial state created by the pump pulse nor any details of the energy transfer dynamics in the 1-exciton subspace. The lack of required microscopic dynamical information is significant, since exact energy transfer dynamics are non-trivial to calculate from first principles.

Here we consider a simple protocol for state tomography, based on an assumed model for calculating the pump-probe response. It is by no means the only such possible state tomography protocol: we choose it because it is straightforward to implement, and turns out to be relatively robust to imperfections such as static disorder. To perform the inversion, we propose to extract an estimate of the excited-state electronic density matrix $\hat{\rho}_e(t)$ from the estimated response function $\hat{R}_{pp}(\omega, \tau)$ at that delay time $\tau$, for each frequency $\omega$, matching the single-exciton transition energies. The relationship between the vector of pump-probe response measurements and the density matrix elements at any fixed time delay is linear [see Eq. (10)], so as long as this map is non-singular, we can solve for the density matrix by simply applying the matrix inverse to the vector formed by these estimated response function points. It is possible that in some circumstances this reconstruction will not yield a valid density matrix, since we did not include the constraint that the reconstruction be positive semi-definite. In this case, then a best estimate to minimize the mean-squared-error of the reconstruction should be obtained using techniques based on maximum likelihood, although we do not encounter this issue for the examples we consider in this paper.

An additional important practical step is the choice of Liouville space in which the extracted state lies. Our results so far hold for transition dipole operators and time evolution without any particular restrictions concerning electronic vs vibrational states or the Hamiltonian we use to describe our system. However, as a practical matter for excitonic energy transfer in light harvesting systems, we are most interested in determining the electronic degree of freedom. The electronic portion of $\rho_{pp}^{(2)}$ has useful structure: namely, it only includes nonzero elements in two blocks, the 0- and 1-excitation subspaces. We denote the projection of the density matrix $\rho$ onto these subspaces by $\rho_0$ and $\rho_1$, respectively. There is only one electronic state in the 0-excitation subspace (the ground state $g$), so the electronic portion of $\rho_0$ must be in that state, $|g\rangle\langle g|$. In the Markov limit, or for delay times much larger than the bath relaxation time, the vibrational portion of $\rho_1$ will be in thermal equilibrium $\rho^{eq}_B$. These facts determine $\rho_0$, up to a
constant of proportionality: the ground state population. Because total probability is conserved in the process of laser excitation, \( \text{Tr} \rho_{pp}^{(2)} = 0 \), so the ground state population is related to the excited state population by \( \text{Tr} \rho_{g}^{(2)} = - \text{Tr} \rho_{e}^{(2)} \). Accordingly, we can write

\[
\rho_{pp}^{(2)} = - \langle g | g \rangle \otimes \rho_{eq}^{B} \text{Tr} \rho_{e}^{(2)} + \rho_{e}^{(2)}. \tag{12}
\]

In this case, the pump-dependence in the pump-probe signal [Eq. (3)] is entirely contained in the excited state portion of the density matrix. Since for weak fields \( \rho_{e} \approx \rho_{eq}^{B} \), with Eqs. (10) and (12) we have a linear map from any excited state density matrix \( \rho_{e} \) to the corresponding pump-probe response. For a system with \( n \) electronic states, we can parameterize this unnormalized density matrix in terms of a linear combination of \( n^{2} \) real parameters, since the excited state density matrix is positive (see Appendix B) and thus Hermitian.

A brief discussion of the scalability of this approach is in order. Based on the real and imaginary parts of the pump-probe response function in the magic angle configuration, our state tomography protocol in principle has \( 2n \) independent real parameters from which to extract the \( n^{2} \) real parameters (including normalization) necessary to describe an arbitrary excited state density matrix of \( n \) electronic states. Accordingly, we cannot necessarily expect this procedure to scale beyond a dimer (\( n = 2 \)), for which we numerically demonstrate the success and stability of this inversion procedure in Sec. IV.

The recently proposed quantum process tomography algorithm based on peak and cross-peak amplitudes in 2D spectroscopy has similar scaling difficulties. It requires determining \( n^{4} - n^{2} \) real parameters in the process matrix from at most \( 12n^{2} \) possible experimental measurements: the real and imaginary signals, \( n \) coherence and \( n \) rephasing frequencies, at most \( 3 \) independent polarization configurations and \( 2 \) phase-matched geometries. These estimates, however, hold only for this specific approach and with a randomly oriented ensemble. Oriented or single molecule measurements offer a much larger number of independent polarization measurements, a point we will return to Sec. V.

**IV. EXAMPLE: DIMER MODEL**

To understand in more detail how the quantum state determines the pump-probe signal, we consider the case of the signal for a dimer of coupled pigments. In general, we can write an effective Hamiltonian for the electronic excited states of a dimer in the form

\[
H_{\text{el}} = E_{1}a_{1}^\dagger a_{1} + E_{2}a_{2}^\dagger a_{2} + J(a_{2}^\dagger a_{1} + a_{1}^\dagger a_{2}). \tag{13}
\]

The terms \( E_{1} \) and \( E_{2} \) are the transition energies of sites 1 and 2, and \( J \) is the pigment-pigment coupling energy. We restrict the system to at most one excitation on each site, so our state space is spanned by the set \( \{|g\}, |e_{1}\}, |e_{2}\}, |f\} \), denoting the ground state, excitation of the first or second site, and excitation of both sites. We further assume the usual linear coupling to a bath of phonons. Details of the bath are specified below. The electronic part of this Hamiltonian can be diagonalized by applying a unitary rotation \( U \) to the single-excitation subspace, given by

\[
U = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}, \tag{14}
\]

where we defined the mixing angle \( \theta = \frac{1}{2} \arctan(2J/\Delta) \) with \( \Delta = E_{1} - E_{2} \). These single excitation eigenstates are denoted \( |\alpha\} \) and \( |\beta\} \). The transition dipole moments for each pigment are \( d_{1} \) and \( d_{2} \), oriented with relative angle \( \phi \).

**A. Analytical calculation of pump-probe response**

To begin, we choose a parametrization for the excited state density matrix of a dimer. In general, any valid density matrix for a two-level system can be written in any basis in form \( \frac{1}{2}(I + \mathbf{r} \cdot \mathbf{\sigma}) \), in terms of the Pauli matrices \( \mathbf{\sigma} = \{\sigma_{x}, \sigma_{y}, \sigma_{z}\} \) and the Bloch vector \( \mathbf{r} = \{r_{1}, r_{2}, r_{3}\} \), with \( r_{i} \) real and \( |r|^{2} \leq 1 \). This can be straightforwardly generalized to unnormalized density matrices by adding the normalization \( r_{0} \) and defining \( r_{0} = 1 \), in which case the set of valid but unnormalized states are those that can be written in the form \( \frac{1}{2}(\mathbf{r} \cdot \mathbf{\sigma}) \), where \( \mathbf{r} \) is now the four-dimensional vector \( \{r_{0}, r_{1}, r_{2}, r_{3}\} \) with constraints \( r_{1}^{2} + r_{2}^{2} + r_{3}^{2} \leq r_{0}^{2} \) and \( r_{0} > 0 \). We will use these four real parameters to parametrize the excited state electronic density matrix \( \rho_{e} \) for our tomography protocol, since it has population \( r_{0} \ll 1 \). Using this representation, the total second-order correction to the electronic density matrix from Eq. (12) is given by

\[
\rho_{pp}^{(2)} = -r_{0}|g\rangle \langle g| + \frac{\mathbf{r} \cdot \mathbf{\sigma}}{2}. \tag{15}
\]

For convenience, we suppose the state is written in terms of the eigenbasis expansion of the excited states \( \{|\alpha\}, |\beta\}\), so the parameters \( r_{1} \) and \( r_{2} \) correspond to excitonic coherences and \( r_{3} \) corresponds to the balance of population between excitonic states. In practice, the experimental signal is only known up to a constant factor, so we can only hope to be able to reliably determine the normalized excited-state density matrix, given by the usual Bloch-vector elements \( \{r_{1}/r_{0}, r_{2}/r_{0}, r_{3}/r_{0}\} \).

It is now straightforward (if tedious) to write down the exact pump-probe response function in terms of microscopic parameters. For illustrative purposes, we do so here for a dimer with a Markovian bath described by Redfield theory in the secular approximation. The time-evolution contained directly in the pump-probe response function is for times following the probe interaction, so the relevant part of the system density matrix for this evolution includes coherences between ground and singly excited states and between singly and doubly excited states. In secular-Redfield theory, coherences in the excitonic basis only evolve with exponential decay, \( G(T)|\alpha\rangle \langle b| = e^{-\gamma_{ab}T} \Theta(T)|\alpha\rangle \langle b| \), where \( G(T) \) is the retarded material Green function denoting evolution for time \( T \), \( \Theta \) is the Heaviside step function, and \( \gamma_{ab} \) is some complex number with positive real part. Since the formulas for the response function will accordingly be most compact in the excitation basis, we consider the excitonic transition dipole moments...
Using Eq. (10), the calculation of the pump-probe response function for an arbitrary state is determined by the vectorized version of the pump-probe operator, \( \langle\langle P(\omega)\rangle\rangle \). For this dimer problem, we define the pump-probe bra vector such that \( R_{pp} = \langle\langle P|\rangle\rangle \), where \( | P \rangle = r = \{r_1, r_2, r_1, r_2\} \). Such a relation still holds upon substitution of \( | P \rangle \) for \( \rho_{pp}^{(2)} \), since the relation between the two given by Eq. (15) is linear. With this convention, evaluating the pump-probe operator in Eq. (9) for this dimer model described by secular-Redfield theory yields the general result,

\[
\langle\langle P \rangle\rangle \propto \begin{bmatrix}
-3\mu_{ga} f_a + \mu_{gf} f'_a - 3\mu_{g\beta} f_\beta + \mu_{af} f'_\beta \\
-\mu_{ga}\mu_{g\beta}(f_a + f_\beta) + \mu_{af}\mu_{f\beta}(f'_a + f'_\beta) \\
-i[\mu_{ga}\mu_{g\beta}(f_a - f_\beta) + \mu_{af}\mu_{f\beta}(f'_a - f'_\beta)] \\
-\mu_{ga} f_a - \mu_{gf} f'_a + \mu_{g\beta} f_\beta + \mu_{af} f'_\beta
\end{bmatrix}^T.
\]

This equation holds for each single molecule that would contribute to the pump-probe signal. We can also calculate the exact isotropic average of Eq. (18) over an ensemble of randomly oriented molecules. In terms of the original Hamiltonian parameters, it is given by

\[
\langle\langle P \rangle\rangle_{iso} \propto \begin{bmatrix}
(cos^2 \theta + \delta^2 \sin^2 \theta)(f'_a - 3f_a) + (\sin^2 \theta + \delta^2 \cos^2 \theta)(f'_\beta - 3f_\beta) + \delta \sin 2\theta \cos \phi(-f'_a + f'_\beta - 3f_a + 3f_\beta) \\
-\frac{1}{2}(\delta^2 - 1) \sin 2\theta(f'_a + f'_\beta + f_a + f_\beta) + \delta \cos 2\theta \cos \phi(f'_a + f'_\beta - f_a - f_\beta) \\
i[\frac{1}{2}(\delta^2 - 1) \sin 2\theta(f'_a - f'_\beta + f_a - f_\beta) + \delta \cos 2\theta \cos \phi(-f'_a + f'_\beta + f_a - f_\beta)] \\
-(\cos^2 \theta + \delta^2 \sin^2 \theta)(f'_a + f_a) + (\sin^2 \theta + \delta^2 \cos^2 \theta)(f'_\beta + f_\beta) + \delta \sin 2\theta \cos \phi(f'_a + f'_\beta - f_a - f_\beta)
\end{bmatrix}^T,
\]

where \( \theta \) is the excitonic mixing angle, \( \delta = |d_2|/|d_1| \) is the ratio of the two site transition dipole moments and \( \phi \) is the angle between them. Note that neither of these equations includes the effects of static disorder, which could be accounted for by averaging the pump-probe response function over each member of the ensemble. Formally, it does not suffice to separately average \( \langle\langle P \rangle\rangle \), since under static disorder the state \( | P \rangle \) also varies in correlated way (see Sec. II B).

Equation (19) makes it possible to determine some cases in which solving for the isotropically averaged state cannot possibly be successful, regardless of the exact inversion protocol. We can identify these cases because successful inversion requires that the elements of \( \langle\langle P \rangle\rangle_{iso} \) be linearly independent. For example, in the homodimer case with both pigments fixed to have the same transition energies (\( \theta = \pi/4 \) or \( \theta = 3\pi/4 \)) and equal transition-dipole moment magnitudes (\( a = 1 \)), the pump-probe signal does not depend on the coherence terms, so it will be impossible to determine them (\( r_1 \) and \( r_2 \)). Likewise, the coherence terms do not contribute if the transition dipole moments have identical magnitude (\( \delta = 1 \)), and either these are oriented perpendicularly (\( \cos \phi = 0 \)) or there are matching dephasing rates for the 0-1 and 1-2 coherences (\( f_a = f'_a \) and \( f_\beta = f'_\beta \), as occurs in the high-temperature limit). As Yuen-Zhou et al. found for the same dimer model, quantum process tomography also fails under similar but not identical conditions.

B. Numerical example

For a numerical example, we consider the dimer model used in a prior investigation of quantum process tomography. We model excitation by a resonant 40 fs full-width-at-half-maximum (FWHM) pump centered at 12 800 cm\(^{-1}\). The parameters in the electronic Hamiltonian are \( E_1 = 12 881 \text{ cm}^{-1} \), \( E_2 = 12 719 \text{ cm}^{-1} \), and \( J = 120 \text{ cm}^{-1} \), and the experiment is performed on an ensemble with normally distributed static disorder of standard deviation 40 cm\(^{-1}\) added to each site energy. The transition dipole moments are fixed with ratio \( \delta = |d_2|/|d_1| = 2 \) and orientation angle \( \phi = 0.3 \). Each pigment is assumed to be coupled to an independent bath of phonons, with spectral density of the form \( J(\omega) = \frac{2\delta}{\omega_c^2} \text{ e}^{-\omega/\omega_c} \), with \( \omega_c = 120 \text{ cm}^{-1} \) and

\[\mu_{ga} = \mu_1 \cos \theta + \mu_2 \sin \theta, \quad (16a)\]
\[\mu_{gb} = -\mu_1 \sin \theta + \mu_2 \cos \theta, \quad (16b)\]
\[\mu_{af} = \mu_1 \sin \theta + \mu_2 \cos \theta, \quad (16c)\]
\[\mu_{bf} = \mu_1 \cos \theta - \mu_2 \sin \theta, \quad (16d)\]

with \( \mu_i = d_i a_i \), where \( d_i \) is the component of the dipole-transition vector parallel to the probe polarization. For convenience, we define \( f_a \) and \( f'_a \) to denote the Fourier transform of the time evolution operator that leads to a peak in the pump-probe spectrum at frequency \( \omega_a \), with decay constant \( \gamma \) or \( \gamma' \), where the prime indicates the decay constant for the transition between the 0- and 1-exciton manifolds:

\[f_a = \frac{1}{i(\omega_a - \omega) - \gamma}, \quad f'_a = \frac{1}{i(\omega_a - \omega) - \gamma'}. \quad (17)\]

We define \( f_\beta \) and \( f'_\beta \) analogously, for the components peaked at \( \omega_\beta \).
\[ \lambda = 30 \text{ cm}^{-1}. \] The bath is assumed to be at thermal equilibrium at \( T = 273 \text{ K} \) and is modeled by secular Redfield theory, including only the real (dissipative) part of the Redfield tensor.

To simulate an experimental dataset, we first calculate the nonlinear polarization \( P(3) (\omega, T) \) for a probe of the same shape as the pump pulse, on a grid of 181 probe frequencies \( \omega \) (intervals of 3.33 cm\(^{-1}\) between 12 500 cm\(^{-1}\) and 13 100 cm\(^{-1}\)) and 140 central time-delays \( T \) between pump and probe pulses (intervals of 6.81 fs between 50 fs and 1 ps). From this nonlinear polarization, we then calculate the results of a hypothetical heterodyne detection with a local oscillator matching the probe pulse, with and without a \( \pi/2 \) phase shift. Finally, we accounted for noise in detection by including additive noise with uniformly random phase and amplitude drawn from a standard deviation with width equal to \( 10^{-2} \) times the maximum amplitude over all delay times and frequencies of the heterodyne detected signal \( S(\omega, T) \). These simulated measurements, generated for comparison both with and without detection noise, are shown in Fig. 1, together the response function from which they are calculated.

C. Response function inversion

Figure 2 illustrates the performance of the completed double Tikhonov regularization based deconvolution algorithm for typical noisy and noise free examples of our test problem. We compare with the “naive” approach of assuming that the probe is impulsive and using Eq. (5) to obtain the response function by simply dividing the signal by absolute value squared of the probe field \( |E_p(\omega)|^2 \). Table I summarizes the results of the simulated inversion for the noise free case and 1000 such noisy examples. In addition to the double Tikhonov and naive methods, we also consider the alternatives of substituting the naive approach individually for each of the two Tikhonov steps. Recall that in the first stage of the inversion \( \lambda \rightarrow P(3) \), the Tikhonov regularization serves

### Table I. Summary of deconvolution performance over 1000 instances of simulated experimental noise. RMSE (root-mean-squared-error) is given by the sum of the absolute difference between the estimated and actual response functions, \( \sqrt{\sum_{\omega, \tau} |\hat{R}_{pp}(\omega, \tau) - R_{pp}(\omega, \tau)|^2} \). Improvement is the multiple of the reduction in RMSE compared to the naive approach. Uncertainties indicate one standard deviation in the empirical distribution.

<table>
<thead>
<tr>
<th>Noise</th>
<th>( \lambda \rightarrow P(3) )</th>
<th>( P(3) \rightarrow R_{pp} )</th>
<th>RMSE</th>
<th>Improvement</th>
</tr>
</thead>
<tbody>
<tr>
<td>( 10^{-2} )</td>
<td>Naive</td>
<td>Naive</td>
<td>12.12 ± 0.07</td>
<td>...</td>
</tr>
<tr>
<td>( 10^{-2} )</td>
<td>Tikhonov</td>
<td>Naive</td>
<td>7.78 ± 0.05</td>
<td>1.6 ± 0.0</td>
</tr>
<tr>
<td>( 10^{-2} )</td>
<td>Naive</td>
<td>Tikhonov</td>
<td>3.34 ± 0.08</td>
<td>3.6 ± 0.1</td>
</tr>
<tr>
<td>( 10^{-2} )</td>
<td>Tikhonov</td>
<td>Tikhonov</td>
<td>0.98 ± 0.06</td>
<td>12.5 ± 0.7</td>
</tr>
<tr>
<td>0</td>
<td>Naive</td>
<td>Naive</td>
<td>7.110</td>
<td>...</td>
</tr>
<tr>
<td>0</td>
<td>Tikhonov</td>
<td>Naive</td>
<td>7.110</td>
<td>1.0</td>
</tr>
<tr>
<td>0</td>
<td>Naive</td>
<td>Tikhonov</td>
<td>0.005</td>
<td>1301</td>
</tr>
<tr>
<td>0</td>
<td>Tikhonov</td>
<td>Tikhonov</td>
<td>0.005</td>
<td>1301</td>
</tr>
</tbody>
</table>
only to smooth the data. It is not surprising then that Table I shows that the specific method chosen for this first stage (i.e., naive or Tikhonov) makes no difference for the noise free case. In the second stage ($P^{(3)} \rightarrow R_{PP}$), the Tikhonov regularization also performs a deconvolution over the probe envelope.

The results in Fig. 2 and Table I show that the Tikhonov based inversion is a clear improvement over the naive approach, reducing the root-mean-squared-error (RMSE) by a factor of 12 for our noisy example and 1300 for our noise-free example. The noisy and noise free examples allow us to observe that the Tikhonov regularizations remove two types of errors inherent in the naive inversion: (1) errors from noisy measurements and (2) errors associated with the convolution of the pump-probe response over the finite probe duration. In the noisy case, both errors are large; in the noise free case, there are only errors from the second source. Clearly, reducing the experimental noise associated with measurement alone does not suffice to accurately estimate the pump-probe response, as the double Tikhonov inversion of the noisy signal outperforms naive inversion of the noise free signal by a factor of 7 in RMSE.

As Figure 2 shows, the errors in the estimate of the response function are not uniformly distributed, revealing structure relevant to our specific example and also to more generic systems. The largest errors are associated with smallest delay times. This makes sense, since the smallest delay times are those at which at the response function (shown in Fig. 1) varies most rapidly. For almost any system, the pump-probe spectrum will change fastest at short delay times, but this is especially true for our example system, where the pump-probe signal includes contributions from quickly oscillating coherences. The Tikhonov estimates face an additional stability challenge at short delays times, since, as discussed above, the reconstruction cannot use measurements from the pulse overlap regime.

### D. State tomography

Since we have demonstrated that the first, response function inversion can be performed with vanishing error, we now consider inverting the exact pump-probe response function to obtain the state of our model dimer. Despite the presence of static disorder in our example, we use the factorization of the response function in Eq. (11). We are obliged to do so even though strictly speaking the relationship does not hold, because the alternative of reconstructing the density matrix for each member of the ensemble from a bulk measurement is unrealistic. Accordingly, even without adding noise associated with the measurement, when carried out for an ensemble, our inversion faces potential stability issues because of the static disorder.

The degeneracies and near degeneracies in Eq. (19) mean that for most Hamiltonian choices our inversion algorithm can only robustly extract at most three of the four parameters necessary to fully characterize the dimer excited state, since the reconstruction matrix will be poorly conditioned. The condition number of a linear transformation gives a bound on the multiplicative increase in the relative error after performing the linear transformation. For our specific numerical example, the condition number drops from 3700 to 3.1 when we include only three parameters. One source of these stability issues for a dimer is evident from Eq. (19): since our numerical example has well separated transition energies, the main contribution to the peaks in the dispersive part of the signal is to the imaginary part of the coherence term, $r_2$. This leaves our inversion to recover three parameters ($r_0$, $r_1$, and $r_3$) from the two peak amplitudes in the absorptive signal. Since recovering three unknowns from two equations is not possible, we need to fix one of these values in order to make the inversion stable. The obvious choice is to fix the normalization $r_0$, since the total excited state population should remain constant after the end of the pump pulse until spontaneous decay, on timescales approaching 1ns for natural pigment-protein complexes. To determine the normalization, we solve for it at a moderately long delay time (e.g., $\tau = 10$ ps) at which point we can safely assume (at least under secular Redfield dynamics) that the real part of the coherence $r_1 \rightarrow 0$, but very few excitations have been lost. If these timescales are not easily separable, then this normalization term could be fit to an exponential decay.

The results of applying this state tomography procedure to our numerical example with varying levels of static disorder are shown in Fig. 3. The fidelity, ranging from 0 to 1, provides a numerical summary of the quality of the state reconstruction. For the level of static disorder chosen by Yuen-Zhou et al. (40 cm$^{-1}$), the reconstruction [panel (a)] has a worst-case fidelity of 99.5% over delay times $T$ shorter than 1 ps, and an average-case fidelity of 99.9%. However, we see that both the worst-case and average-case fidelities drop sharply as the static disorder is increased above this level [panel (b)], since our assumption that the pump-probe response can be factorized between the pump-probe projection and the second order density matrix becomes increasingly unrealistic.

### V. SCALING TO LARGER SYSTEMS

Can state tomography scale to systems larger than an excitonic dimer? In particular, can we apply it to precisely reveal the excitonic state in a natural light-harvesting system? Any scaling difficulties will be encountered in the second step of our inversion protocol, to determine the quantum state from the response function, since the relationship between the response function and measured signal does not directly depend on system parameters. Successful state tomography certainly requires both knowledge of how each density matrix element contributes to measurements and appropriate conditions such that each element, at least in principle, makes an independent and non-zero contribution. By construction, these conditions were satisfied for our hypothetical dimer example. We found that the primary limitation on solving for the state was ensemble disorder, which can in principle be avoided by single molecule techniques. Now, to explore the limits of state tomography, we relax these assumptions in order to consider the feasibility of state tomography in an actual protein-pigment complex.
As a model light-conversion system, we focus on a monomer of the Fenna-Matthews-Olson (FMO) complex of green sulfur bacteria, which consists of 7 pigment molecules. The FMO complex is a widely used model system for understanding photosynthetic energy transfer and is thus one of the best characterized natural protein-pigment complex. The crystal structure for the FMO complex is known, which combined with input from spectroscopy experiments, has allowed for general agreement on an electronic Hamiltonian. Because the arrangement of pigments is fairly disordered, each excited state in a monomer of the FMO complex is bright, although they overlap in the presence of homogeneous and inhomogeneous broadening. This is important, since full state tomography would certainly not be possible on a system with multiple dark states, because no optical probes could reveal the distribution of energy among those states. However, typical of the situation for other natural pigment-protein systems, there is little consensus on the magnitude of the static disorder or the spectral density of the electronic-vibrational coupling. These difficulties are compounded by the theoretical and computational challenge of modeling dynamics in a system as large as FMO exactly for arbitrary system-bath interaction strength. For our concrete example, we use the electronic Hamiltonian for FMO of Chlorobaculum tepidum from Ref. with the spectral density and computational model of secular Redfield theory matching those used in for the dimer example. This model includes only one electronic state per pigment and assumes Gaussian distributed static disorder with standard deviation 42.5 cm\(^{-1}\) (100 cm\(^{-1}\) FWHM).

Our formalism for the pump-probe response function allows us to place bounds on the feasibility of any state tomography procedure, since the relationship between system information and the resulting pump-probe spectra is entirely contained in the pump-probe response operator. By looking at the ensemble average of this operator, we implicitly consider inversion under the scenario that the average over static disorder can be factorized between the pump-probe operator and the density matrix. This assumption was successful in the dimer example above when the magnitude of static disorder was not too large. As discussed in Sec. II B, the pump-probe operator at each frequency can be interpreted as a Liouville space bra-vector \(\langle P(\omega)\rangle\). Accordingly, it is possible to interpret the calculation of a pump-probe response as the act of applying the linear operator

\[
\mathbb{P} = \int d\omega \langle \rho^{(2)}(\omega) \rangle
\]

(20)

to the state \(|\rho^{(2)}_{PP}\rangle\). We now consider the properties of the linear operator \(\mathbb{P}\) in the limit of effectively continuous sampling of probe frequencies \(\omega\). To represent states in Liouville space, we use a basis set that allows us to represent each state with \(n^2\) real values, in terms of populations \(|n\rangle\langle n|\) and coherences \(|m\rangle\langle n|\) and \(i|n\rangle\langle m| - i|m\rangle\langle n|\). This allows us to construct a real-valued version of the map \(\mathbb{P}\) that takes real valued state vectors to real valued spectra by concatenating the real and imaginary parts of \(\mathbb{P}\).

To begin, in Figure 4 we plot the elements of the absorptive (real) part of the pump-probe operator for our model of the FMO complex in the isotropic average (magic angle configuration). We represent the operator in terms of the species associated spectra of each excitation population and the real and imaginary part of each coherence, so that the pump-probe spectra of any excited state is equal to the linear combination of the plotted spectra weighted by the indicated density matrix elements. In addition to the unperturbed spectra, we also plot the range of possible spectra given current uncertainty about the best fit parameters. We conservatively estimate the uncertainty in the electronic Hamiltonian by sampling over additive independent Gaussian noise of width 20 cm\(^{-1}\) for each site energy and 10% of the value of each off-diagonal coupling. This uncertainty is in addition to the static disorder, which we leave with fixed magnitude. The most striking feature of these spectra is that, at least in the isotropic average, the dominant contribution to the pump-probe spectra is from the population terms. The smaller contribution of most coherence terms, compounded by the already smaller values of the coherences in the density matrix due to dephasing, explains why it is difficult to observe oscillations due to electronic coherence in pump-probe spectra. Even for extremely
FIG. 4. Species associated spectra, defined by the contribution of the marked density matrix elements to the pump-probe response, for the FMO complex at 77 K (blue) and 300 K (red), obtained as the average of 1000 samplings over static disorder. Labels indicate the contributing density matrix element in the excitonic basis. Shaded regions indicate central 95% confidence intervals obtained from 1000 additional samplings over Hamiltonian uncertainty, as described in the text.

precise measurements, the uncertainty in some of these species associated spectra suggests that our current Hamiltonian characterization does not suffice to reliably obtain most density matrix elements. Indeed, the dominance of the diagonal terms suggests that a practical scheme for partial state tomography could consist of entirely ignoring the off-diagonal terms.

Another approach to estimating the feasibility of inversion for arbitrary states is to look at the spectral properties of the operator $P$ as revealed by the singular value decomposition, $P = USV^†$, where $U$ and $V$ are unitary and $S$ is diagonal with positive elements. In particular, we focus on the singular values $\sigma_i$, given by the diagonal elements of $S$ in descending order and normalized to the highest singular value $\sigma_1$. To compare the feasibility of inversion under various conditions, we plot these singular values in Figure 5. The singular values reveal significant information about the feasibility of an inversion: in general, inversion is more feasible when the singular values $\sigma_i$ decay more slowly. For example, the condition number, which gives an upper bound on the ratio by
which the relative error can increase in an inversion, is equal
to the ratio of the largest to the smallest singular values. In
Figure 5, the condition number is one over the value shown for
$i = 49$.

Figure 5 provides an indication of the relative signifi-
cance of different experimental constraints insofar as they af-
fect state tomography. We see two major changes that reduce
the condition number of the inversion by around two orders
of magnitude each. First, reducing temperature from 300 K to
77 K improves the conditioning because spectral features in
the species associated spectra are sharpened. Second, chang-
ing from a randomly oriented (isotropic) to an oriented sam-
ple helps because in principle 9 times more independent mea-
surements are possible than in the magic angle configuration,
each independent polarization configuration of the probe and local oscilla-
tor, including the ensemble average over static disorder. The labeling of each
species matches that used in Figure 4: all entries including and above the
diagonal correspond to the real part of the matching density matrix element
(in the excitonic basis), and all entries below the diagonal correspond to
the imaginary part. The cartesian coordinates were chosen arbitrarily, matching
those used in an assignment of the crystal structure.

VI. CONCLUSIONS

What information does an ultrafast spectroscopy exper-
iment tell us about an excitonic system? How can we best
design these experiments? We believe that a powerful way to
answer these questions is to treat ultrafast spectroscopy as a
hierarchy of statistical inverse problems, as we have demon-
strated here for estimating the electronic excited state. By
dividing our analysis into many smaller steps—from experi-
mental signal to response function to excited state and even-
tually on to dynamics, equations of motion and Hamiltonian
parameters—we can see exactly how and where our mod-
els and experiments fall short. In this regard, our approach
contrasts strongly with the established procedure of “for-
ward simulation” for determining Hamiltonian parameters in
complex excitonic systems by simultaneously fitting many
experiments with an assumed theoretical model for calculat-
ing spectra from first principles.36,38 Moreover, to identify the
time-evolving excited state or another intermediate quantity
in our approach, we do not need to introduce additional errors
by recalculating from first principles.

Our multi-stage inversion has clear extensions to more
general nonlinear spectroscopies beyond QST and pump-
probe. As we describe in Appendix E, we could apply es-
sentially the same inversion procedure to a photon-echo ex-
periment to determine the phase-matched component of the
2nd-order density matrix that contributes to the observed sig-
nal [Eq. (E1)]. If it is possible to construct a full set of inde-
dendent phase-matched initial conditions, our state tomogra-
phy procedure could be used for process tomography, along
the lines of a previous proposal.5 More generally, this sug-
gests a new indirect approach for process tomography: first,
field information should be used to obtain the 3rd-order re-
sponse function; second, system information should be used
to extract the process matrix. Determining the response func-
tion as an intermediate quantity allows us to be sure we have
obtained the maximum information from experiments before

![Normalized singular values from a singular value decomposition of the real valued pump-probe map $P$ given by Eq. (20) for the FMO complex under various conditions. The top line is from the spectra of a single monomer at 77 K, from the combination of measurements in all independent polarization configuration. Subsequent lines add additional constraints, which apply cumulatively: ensemble measurement (over static disorder), the isotropic average of the signal, only the absorptive (real) part of the signal and finally performing the measurement at room temperature.](image)

![Maximum amplitude over probe frequencies of the species associated spectra for a FMO monomer at 77 K for (a) the isotropic average and (b) each independent polarization configuration of the probe and local oscillator, including the ensemble average over static disorder.](image)
considering the harder theoretical problem of extracting system parameters, such as the state, process matrix or underlying Hamiltonian. Such extensions will be pursued in future work.

More immediately, our approach is particularly well suited to evaluating the benefits of employing colored pulses or an ultrafast pulse shaper, because our formalism makes no assumptions about the shape of pump and probe pulses: they are only required not to overlap in time. In contrast, prior assumptions about the shape of pump and probe pulses: they are suited to evaluating the benefits of employing colored pulses work.

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**ACKNOWLEDGMENTS**

In a pump-probe experiment, the first two interactions are with the same pulse \((k_1 = k_2)\), called the pump, and the signal is observed in the direction \(k_3 = k_3\), so \(u_1 = -u_2\). The third interaction is with the probe field. Accordingly, the signal is given by adding together the rephasing and non-rephasing interactions, and both pulse orderings 1-2-3 and 2-1-3. Re-arranging the terms that result from inserting Eq. (A2) into Eq. (A1) yields Eqs. (1)–(3).

**APPENDIX B: POSITIVITY OF \(\rho^{(2)}\)**

We consider the system density matrix \(\rho\) in the presence of weak fields of strength \(\epsilon \ll 1\). Let \(\rho^{(n)}\) denote the contribution to the density matrix of strength \(O(\epsilon^n)\). Thus we can write \(\rho = \rho^{(0)} + \rho^{(1)} + \rho^{(2)} + \rho^{(3)} + O(\epsilon^4)\). Since the temperature is typically several orders of magnitude smaller than the electronic energy gap, the system starts in a tensor product of the electronic ground state \(|g\rangle \langle g|\) and the equilibrium vibrational state \(\rho^{(0)} = |g\rangle \langle g| \otimes \rho^{(0)}_{\text{eq}}\).

We are interested in the excited state portion of \(\rho^{(2)}_{\text{eq}}\), the component of \(\rho^{(2)}\) that contributes to the phase-matched signal \(k_3 = k_3\) given by Eq. (3). We write this excited state density portion as \(\rho^{(2)} = Q\rho^{(2)}\), where \(Q\) denotes the projection onto the 1-excitation manifold. Since the non-phase-matched components involve 2-excitation states, we also have \(\rho^{(2)} = Q\rho^{(2)}\). This projection is given by \(Q\rho = I_1\rho I_1^\dagger\), where \(I_1 = \sum_m |m\rangle \langle m|\) is the identity operator restricted to the 1-excitation manifold and \(|m\rangle\) denotes the state where only pigment \(m\) is excited. Since the map \(Q\) is written in the appropriate form and \(\sum_m I_1^\dagger I_1 = I\), \(Q\) is completely positive, with \(0 \leq \text{Tr} Q \rho \leq 1\).

Moving from the ground state to the 1-excitation manifold requires at least two applications of the creation/annihilation operators contained in the dipole transition operators \(\mu^{(2)}\), and a dipole operator must be applied an even number of times. Accordingly, \(Q\rho^{(n)} = 0\) for \(n = 0, 1, 3\), which leaves \(Q\rho = \rho^{(2)} + O(\epsilon^4)\). Since we proved \(Q\rho\) is positive and \(\rho^{(2)}\) itself is \(O(\epsilon^2)\), we have shown that \(\rho^{(2)}\) is positive, up to relative errors of \(O(\epsilon^2)\). Any positive operator is a valid density matrix if it has trace one.
APPENDIX C: TIKHONOV REGULARIZATION

The convolution of the pump-probe response with the probe pulse in Eq. (4) and the nonlinear polarization with the local-oscillator in Eq. (6) are both particular cases of a Fredholm integral equation of the first kind. An extensive literature exists on numerical inversion of such equations, known as discrete inversion problems.\(^{32}\) In general, the discretization of such an integral equation can be written as

\[
b = Ax + e, \quad (C1)
\]

where \(b\) is the measured signal [e.g., the nonlinear polarization \(P^3(\omega, T)\), \(x\) is the desired quantity to obtain [e.g., \(R_{PP}(\omega, T)\)], \(A\) is a linear operator representing the integral equation with appropriate coefficients (determined here by the probe field \(E_p\)) and \(e\) denotes some additive experimental error inherent in the data collection. The obvious solution to estimating \(x\) from \(A\) and \(b\) is to calculate \(\hat{x} = A^{-1}b\). However, in practice \(A\) may not be invertible. This is the case for our inversion, since we ignore the experimental signal for times in the pulse overlap regime but still attempt to reconstruct the pump-probe response at those times, which guarantees that \(b\) has a lower dimensionality than \(x\). In addition, the presence of even a vanishingly small amount of experimental noise \(e\) makes exact least-squares minimization unsuitable; it will over-fit the noise component \(e\).

Accordingly, to calculate a robust estimate \(\hat{x}\) of \(x\) we use the general form of Tikhonov regularization,\(^{32}\)

\[
\hat{x} = \arg \min_x \left\{ \|Ax - b\|^2 + \lambda^2 \|Lx\|^2 \right\}. \quad (C2)
\]

Tikhonov regularization can be derived formally from the perspective of Bayesian inference, given normally distributed errors and priors.\(^{34}\) It can be equivalently expressed as the linear least-squares problem, \(\min \|Lx - b\|^2\), and thus the exact solution is given by \(\hat{x} = \left( L^T L + \lambda^2 D \right)^{-1} L^T b\), where \(D\) denotes the Penrose-Moore pseudoinverse.\(^{32}\) Ideally, the linear operator \(L\) and (real) regularization parameter \(\lambda\) are chosen so that \(\lambda^2 \|Lx\|^2\) is an optimally weighted penalty on undesirable features of the solution \(x\), reflecting our prior knowledge of the general form of \(x\). Common choices of \(L\) include the identity matrix \(I\) and finite-difference approximations to the first or second derivative given by \((D_1x)_n = x_n - x_{n-1}\) and \((D_2x)_n = x_{n+1} - 2x_n + x_{n-1}\). We compare different techniques for selecting \(\lambda\) and \(L\) in Appendix D.

There are a number of powerful techniques for calculating efficient approximate solutions to Eq. (C2), especially in cases where the linear operator \(A\) is structured,\(^{35}\) such as in our case, where \(A\) is a Toeplitz matrix. For several hundred time-delays or probe frequencies, we find that we can solve each deconvolution with Eq. (C2) exactly and quickly (\(~1\) s on a modern CPU) by calculating the singular value decomposition of the matrix \([\lambda L]\). In principle, it would be possible to solve both steps in the inversion of the pump-probe response in a single two-dimensional Tikhonov regularization. Such 2D inversions are routinely performed in image processing,\(^{32}\) but would require slower, more approximate techniques than the exact solution we used here. Since we find significant improvement without invoking these more complicated methods, we do not use them here.

APPENDIX D: PARAMETER SELECTION FOR TIKHONOV REGULARIZATION

To perform deconvolutions using Eq. (C2), we need to choose a procedure to select the regularization parameters \(\lambda\) and \(L\). In practice, there are a wide variety of techniques for making these selections and the best choice depends on the particular problem at hand.\(^{32}\) Here we compare the performance of different techniques on simulations matching the dimer problem we analyze in Sec. IV.

To begin, we compared the performance of general form Tikhonov regularization with \(L\) equal to \(I\), \(D_1\), and \(D_2\), with \(\lambda\) chosen optimally so as to minimize the exact mean-squared-error \(\|\hat{x} - x\|^2\). A summary of reconstructions of \(R_{PP}(\omega, \rho_{PP}(T))\) for 1000 instances of low and high noise is shown in Table II. As a benchmark, we consider the ratio of the mean-squared-error from the Tikhonov estimate to the mean-squared-error of our naive estimate \(\hat{R}_{PP}(\omega, \rho_{PP}(T)) \propto P(\omega, T)/E_{pp}(\omega)\), which holds in the limit of an instantaneous probe pulse [Eq. (5)]. For our state inversion algorithm, reconstruction of the response function is most important at frequencies matching the exciton transition energies, so we picked \(\omega = \omega_a\), the transition frequency of the higher energy exciton state. We found qualitatively similar results for \(\omega = \omega_g\) and other choices of \(\omega\) as well. As Table II shows, with exact selection of the best regularization parameter \(\lambda\), we found consistently best performance with \(D_2\), the linear operator approximating the second derivative of the response function with respect to the delay time \(T\). This is an intuitively reasonable choice, since plausible response functions should be smooth.

With the choice for \(L\) determined, we also need a realistic procedure for selecting the regularization parameter \(\lambda\). In a true inversion problem the response function \(x\) is unknown, so we cannot choose \(\lambda\) to minimize the exact mean-squared-error. There are a variety of standard techniques for making this choice, with performance that can vary widely depending

<table>
<thead>
<tr>
<th>Noise</th>
<th>Penalty</th>
<th>Selection</th>
<th>(\lambda_{\text{opt}})</th>
<th>Improvement</th>
</tr>
</thead>
<tbody>
<tr>
<td>(10^{-2})</td>
<td>(I)</td>
<td>Exact</td>
<td>0.151 ± 0.007</td>
<td>2.3 ± 0.2</td>
</tr>
<tr>
<td>(10^{-2})</td>
<td>(D_1)</td>
<td>Exact</td>
<td>0.340 ± 0.034</td>
<td>4.9 ± 0.7</td>
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<tr>
<td>(10^{-2})</td>
<td>(D_2)</td>
<td>Exact</td>
<td>0.67 ± 0.11</td>
<td>6.2 ± 1.2</td>
</tr>
<tr>
<td>(10^{-2})</td>
<td>(D_2)</td>
<td>GCV</td>
<td>0.78 ± 0.17</td>
<td>5.7 ± 1.2</td>
</tr>
<tr>
<td>(10^{-2})</td>
<td>(D_2)</td>
<td>NCP</td>
<td>2.57 ± 0.36</td>
<td>2.3 ± 0.4</td>
</tr>
<tr>
<td>(10^{-3})</td>
<td>(I)</td>
<td>Exact</td>
<td>0.049 ± 0.017</td>
<td>5.9 ± 0.5</td>
</tr>
<tr>
<td>(10^{-3})</td>
<td>(D_1)</td>
<td>Exact</td>
<td>0.073 ± 0.010</td>
<td>30 ± 5</td>
</tr>
<tr>
<td>(10^{-3})</td>
<td>(D_2)</td>
<td>Exact</td>
<td>0.112 ± 0.017</td>
<td>56 ± 12</td>
</tr>
<tr>
<td>(10^{-3})</td>
<td>(D_2)</td>
<td>GCV</td>
<td>0.103 ± 0.019</td>
<td>52 ± 12</td>
</tr>
<tr>
<td>(10^{-3})</td>
<td>(D_2)</td>
<td>NCP</td>
<td>0.449 ± 0.039</td>
<td>19 ± 2</td>
</tr>
</tbody>
</table>
on the problem being solved, so selection of an appropriate method requires more tests on simulated data. We considered two such methods for $L = D_2$: generalized cross-validation (GCV) and the normalized periodogram (NCP). We calculate the GCV error using the exact pseudoinverse solution and the NCP error by adding together the errors for the real and imaginary parts of the spectra. We then minimize these error estimates as a function of $\lambda$ using a one-dimensional search with the downhill simplex method. We also impose the additional restriction $\lambda \geq 5 \times 10^{-11}$ to avoid convergence failures with our SVD implementation that we encountered when performing deconvolutions on noise-free simulated spectra. The results, also shown in Table II, show that GCV is the best choice for our test problem, with performance nearly matching that of exact selection technique. In contrast, NCP systematically overestimated the noise, as indicated by regularization parameters about four times larger than the exact selection method. However, NCP still offered an improvement in the reduction of the mean-squared-error compared to the naive approach.

**APPENDIX E: ALTERNATIVE FORMULATION**

There are several obvious extensions or alternatives to the recipe described in Sec. II. For example, the exact same relations in Eqs. (1)–(3) hold for a general photon-echo (or non-rephasing) experiment with two distinct pump pulses, except that in this situation the sum in Eq. (3) to determine the portion of $\rho^{(2)}$ that contributes to the signal should be removed to leave only one of the two phase-matched contributions. Instead, the photon-echo (PE) signal depends on the second-order density matrix given by

$$\rho_{PE}^{(2)}(t) = \left(\frac{i}{4}\right)^2 \int_0^\infty dt_2 dt_1 G(t_2) V^{(1)} G(t_1) V^{(1)} \rho_0 \times E_1^*(t - t_2) E_1^*(t - t_2 - t_1).$$  \hspace{1cm} (E1)

However, unlike the case for the state $\rho_{pp}^{(2)}$ that contributes to the pump-probe signal, the excited state portion of $\rho_{PE}^{(2)}$ is not necessarily either Hermitian or equivalent to the total excited state density matrix. This makes its interpretation less clear but presents no additional technical difficulties for our inversion procedure.