

Convergence of Time-Derivative Nonadiabatic Couplings in Plane-Wave DFT Calculations

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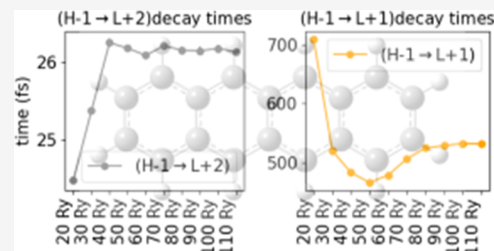
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ABSTRACT: Accurate prediction of charge carrier relaxation rates is essential to design molecules and materials with the desired photochemical properties for applications like photocatalysis and solar energy conversion. Nonadiabatic molecular dynamics allows one to simulate the relaxation process of excited charge carriers. Plane-wave density functional theory (DFT) calculations make the time-derivative nonadiabatic couplings (TNACs) simple to compute because the basis is independent of the atomic positions. However, the effect of the kinetic energy cutoff for the plane-wave basis on the accuracy of the dynamics has not been studied. Here, we examine the effect of the kinetic energy cutoff on the TNACs and decay time scales for the prototypical model system of tetracene.

These calculations show that the choice of kinetic energy cutoff can change the relaxation time by up to 30%. The relaxation times of states that have small TNACs to other states or are far from degenerate are more sensitive to the kinetic energy cutoff than those of states with large TNACs or near degeneracies. A kinetic energy cutoff of 60 Ry is sufficient for all states to reach semiquantitative agreement (absolute error <10%) with the decay times of our 110 Ry reference data, and a cutoff of 80 Ry is required for all states to reach quantitative agreement (absolute error <2%).



1. INTRODUCTION

Understanding the photophysical properties of molecules^{1–3} is essential to controlling important properties such as competition between radiative or nonradiative relaxation pathways,⁴ photocatalytic activity,⁵ and photoinduced charge separation.^{6,7} These properties are dependent on the coupled nuclear and electronic dynamics following excitation by light; thus, an accurate prediction of these dynamics is essential. Nonadiabatic molecular dynamics (NAMD) simulations describe the coupled evolution of the electronic and nuclear degrees of freedom during nonradiative relaxation of a photoexcited molecule.^{8,9} The fewest switches surface hopping (FSSH) approach is a common method for simulating NAMD.^{8–10} In the FSSH algorithm, the nuclei follow a classical trajectory while the electrons are treated quantum mechanically; as the simulation proceeds, the electrons switch between electronic states with probabilities determined by the coupling between the electronic and nuclear degrees of freedom.¹¹ Details and extensions of the FSSH algorithm have been described thoroughly in many other papers.^{12–17} The nonadiabatic coupling (NAC) vector used to compute the transition probability between states and how to scale atomic velocities after a state transition is computed as

$$d_{ij}(R) = \langle \psi_i(r; R) | \nabla_R | \psi_j(r; R) \rangle \quad (1)$$

where $d_{ij}(R)$ is the NAC between electronic states $\psi_i(r; R)$ and $\psi_j(r; R)$, which depends parametrically on the positions R of the nuclei, and ∇_R is the gradient of the nuclear positions.

FSSH simulations are computationally demanding even for small systems and are only feasible for large or condensed-phase systems with additional approximations that reduce the computational cost.¹⁰ First, approximations are often made to the nuclear dynamics. If the ground-state and excited-state potential energy surfaces are similar, the classical path approximation can be used without substantially changing the nuclear dynamics.^{9,11,18,19} In the classical path approximation, the back-reaction of the excited electrons on the nuclei is neglected and a classical nuclear trajectory that is independent of the electron dynamics is used. Thus, a ground-state trajectory can be precomputed, and the NACs can be precomputed based on a series of known geometries. This approximation is often suitable for the excited-state dynamics of systems with up to hundreds of atoms where an electronic excitation does not significantly change the overall electron density,¹⁸ including molecular complexes,¹⁸ solids,⁸ and condensed-phase systems.¹¹

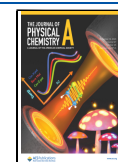
In addition, approximations must be made to the electronic states. Although multireference or configuration-interaction-based approaches typically produce the most accurate

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NACs,^{20,21} they are often too computationally expensive. For large systems, time-dependent density functional theory (TDDFT) may also be too computationally expensive to be tractable for a large number of time steps.¹⁰ Approaches based on time-dependent Kohn–Sham (TD-KS) theory also have a reduced computational cost.^{10,22} TD-KS theory uses the single-particle approximation to model electronic states either as single KS orbitals or as Slater determinants (SDs) involving an excitation from one KS orbital to another.^{11,18,23} This approach has produced results in agreement with TDDFT for systems where the single-particle approximation is reasonable, including organic–inorganic complexes, quantum dots, and some small molecules.¹⁸

When using the classical path approximation and TD-KS orbitals, the formula for the NACs²⁴ can be simplified to the time-derivative NACs (TNACs) between the electronic states at consecutive time steps using the Hammes–Schiffer²⁵ formulation

$$d_{ij} = \frac{\langle \phi_i(t) | \dot{\phi}_j(t + \Delta t) \rangle - \langle \phi_i(t + \Delta t) | \dot{\phi}_j(t) \rangle}{2 \cdot \Delta t} \quad (2)$$

where ϕ_i and ϕ_j are KS orbitals at consecutive time points t and $t + \Delta t$. In the SD basis, the time overlaps are computed using the Lowdin formula^{26,27}

$$\langle \phi_p(t) | \phi_q(t + \Delta t) \rangle = \det S[pq] \quad (3)$$

where $S[pq]$ is the matrix of the time overlaps between the occupied molecular orbitals (MOs) in the SDs $\phi_p(t)$ and $\phi_q(t + \Delta t)$.

Using these approximations to compute the TNACs, several factors can strongly affect the accuracy of the TNACs and thus of the decay time scales. The quality of the computed TNACs depends strongly on the level of theory.^{11,28} For example, generalized gradient approximation (GGA) functionals are well-known to underestimate band gaps; thus, it is unsurprising that the GGA functional PBE predicts TNACs an order of magnitude larger than those computed using hybrid functionals with exact exchange.^{29,30} The accuracy of PBE can be improved by using an electron self-energy correction to correct the band gap.³¹ Aside from the functional used to compute the TNACs, the length of the trajectory and the average magnitude of the TNACs over the trajectory can also affect the accuracy of the decay time scales.³² In addition, decoherence corrections tend to slow excited state dynamics, particularly transfer between energetically distant states. If decoherence corrections are considered, it is possible to model accurate dynamics while only considering the TNACs for a small number of neighboring states.^{33,34}

A factor that has been largely neglected in understanding the accuracy of TNACs is the role of the basis set. TNACs are often computed using plane-wave basis sets because of the simplicity of computing overlaps of these functions; in these calculations, the MOs are described by a plane-wave expansion³⁵

$$\psi_{i,k}(r) = \sum G \psi_{i,k}(G) \frac{e^{i(k+G)r}}{\sqrt{\Omega}} \quad (4)$$

where i is the KS orbital index, k is the Bloch vector, G is the reciprocal lattice vector, and Ω is the volume of the periodic box used to describe our system. Typically, all plane waves with energies below a given kinetic energy cutoff are included in this expansion; the kinetic energy cutoff is often chosen by

converging the total energy of the system with respect to the kinetic energy cutoff. Since a higher kinetic energy cutoff yields a more accurate description of the MOs, the TNACs will depend on the kinetic energy cutoff. However, to our knowledge, it is not known whether the cutoff that converges total energy is sufficient to provide converged TNACs. In previous studies, the kinetic energy cutoff is often provided,^{11,36–44} while other studies report using a converged plane-wave basis.^{18,45–47}

Here, we examine the dependence of the TNACs and excited-state dynamics on the kinetic energy cutoff by performing FSSH simulations using the classical path approximation with TD-KS electronic states. We use tetracene as a prototypical model system: its rigid structure and relatively constant electron density upon excitation make it suitable for the classical path approximation, and its electronic structure is well-known.^{48,49} Our results show that the TNACs converge semiquantitatively (absolute error < 10%) with our reference TNACs at a kinetic energy cutoff of 60 Ry. To achieve quantitative (absolute error < 2%) agreement with our reference data, a kinetic energy cutoff of 80 Ry is required. Our results show that the decay dynamics are much less sensitive to the kinetic energy cutoff than to previously studied factors like the choice of DFT functional.

2. COMPUTATIONAL METHODS

The geometry of tetracene was optimized at the PBE level⁵⁰ within a box with dimensions $4.5 \times 4.5 \times 4.5$ nm and a kinetic energy cutoff of 40 Ry. Starting from this geometry, the ground-state nuclear trajectory was computed at the same level of theory and kinetic energy cutoff for 2400 steps with a time step of 1 fs, resulting in a 2.4 ps trajectory. The Verlet algorithm was used to integrate Newton's equations of motion, and the atomic velocities were adjusted with an Anderson thermostat, coupling our system to a 300 K thermal bath.

To ensure that the FSSH simulations were performed using equilibrated geometries, the TNACs were computed using only time steps from step 300 until the end of the nuclear trajectory. Examination of the temperature of the system (Figure S27) shows that the temperature fluctuations are consistent in magnitude after this time; since the system is a single molecule, the magnitudes of the temperature fluctuations are relatively large. For each time step in this range, the electronic structure was computed using the PBE functional⁵¹ with kinetic energy cutoffs ranging from 20 to 110 Ry in steps of 10 Ry. All calculations were performed using norm-conserving pseudopotentials⁵² within the Quantum Espresso software package.⁵³

The TNACs were computed for all pairs of Slater determinants within a basis set from HOMO – 2 to LUMO + 2, using phase corrections to ensure consistency of the MOs at different time steps.⁵⁴ The excited-state dynamics were computed using the FSSH algorithm with the transition probabilities rescaled by a Boltzmann factor.¹¹ Following the truncation of the trajectory, an initial starting point for the NAMD trajectory was chosen every 20 time steps from 300 to 1900 fs time step for a total of 80 starting points and propagated for 500 fs, and 2500 stochastic trajectories were computed for each starting point. The 1600 fs window for the starting points is large enough that there are four non-overlapping time windows for the NAMD trajectories. Although the spacing between start times is short, we will show in the results that many of the initial states have 20–50%

of their populations transfer out of the initial state within the first 20 fs. We chose to compute 2500 electron trajectories, at least 10 times what is typical in NAMD studies, for each starting point to average out the stochastic contribution to provide a precise estimate of the decay time scales based on the TNACs. The population of the system remaining in the initial state was fit to a single exponential decay

$$f(t) = ae^{-\lambda_i t} \quad (5)$$

where α is the initial population of state i and λ_i is the decay rate. Based on this fit, the relaxation time was computed as $\tau_i = \frac{1}{\lambda_i}$. To confirm that the stochastic contributions to the relaxation times are negligible, the standard deviation (σ) of the relaxation times from the 5 FSSH simulations is calculated as

$$\sigma = \sqrt{\frac{\sum (\tau_i - \bar{\tau})^2}{5}} \quad (6)$$

where τ_i is the relaxation time of the initial state from the i th simulation and $\bar{\tau}$ is the average relaxation time from the five FSSH simulations. The percent error for the relaxation times computed with lower kinetic energy cutoffs are computed as

$$\% \text{ error} = \frac{\bar{\tau}_k - \bar{\tau}_{110\text{Ry}}}{\bar{\tau}_{110\text{Ry}}} \times 100\% \quad (7)$$

where $\bar{\tau}_k$ is the average relaxation time using kinetic energy cutoff k and $\bar{\tau}_{110\text{Ry}}$ is the average of the reference relaxation times computed with a 110 Ry kinetic energy cutoff. TNACs and FSSH simulations were performed using the methodology development code Libra.⁵⁵

3. RESULTS AND DISCUSSION

The overall goal of this study is to examine the effect of the kinetic energy cutoff on TNACs and decay time scales by comparing the results for a series of kinetic energy cutoffs to reference data computed with a cutoff of 110 Ry. We selected tetracene (Figure 1) as a model system because its rigid

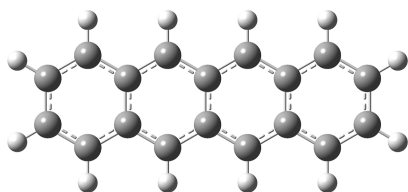


Figure 1. Chemical structure of tetracene.

structure makes the classical path approximation a suitable choice. The optical properties and excited-state dynamics of tetracene have been reported in several studies of polycyclic aromatic hydrocarbons (PAHs).^{48,49}

We focus on an active space from HOMO $- 2$ to LUMO $+ 2$ KS MOs. Within this range, tetracene has two pairs of nearly degenerate MOs: (1) HOMO $- 1$ and HOMO $- 2$ and (2) LUMO $+ 1$ and LUMO $+ 2$. Using the MOs within this range, we generate a SD basis including all possible single excitations of the alpha electrons plus the ground state. The energies of the SDs are calculated as the sum of the energies of the occupied KS in each SD as implemented in Libra

$$E_{\text{SD}_{\text{occ}}} = \sum_{i \in \text{occ}_{\text{KS}}} \epsilon_i \quad (8)$$

where ϵ_i is the orbital energy. The near degeneracy in the MOs results in two groups of SDs with nearly degenerate energies (Figure 2b). The first set is around -20.7 eV, approximately 3 eV above the ground-state energy, and the second set is around -19.2 eV, around 4.5 eV above the ground state. The nearly degenerate pairs of SDs have TNACs that are on average much larger than the TNACs between any nondegenerate pair of SDs (Figure 2a). As expected, the energies of the SDs fluctuate but do not dramatically change throughout the course of the 2400 fs ground-state trajectory.

We now examine the effect of the kinetic energy cutoff. As discussed in the Introduction, convergence of total energy is commonly used as a proxy for convergence of the dynamics. For this system, the total energy of the optimized geometry follows the expected exponential-like decrease with increasing kinetic energy cutoff, converging to within 0.03 Ry of its reference value (110 Ry cutoff) at a kinetic energy cutoff of 100 Ry (Figure 3a). However, this higher accuracy comes at a trade-off of higher computational cost: as the cutoff increases from 20 to 110 Ry, the CPU time for a single-point energy calculation on a single geometry increases from 5 to 35 min as shown in Figure 3b. Since a single geometric trajectory long enough that the NAMD simulations typically contains thousands of geometries, small differences in computational time for each geometry add up over the course of a trajectory.

3.1. TNACs in SD Basis. Although checking convergence of the total energy and electronic states with respect to the kinetic energy cutoff is relatively common in NAMD studies,^{29,56,57} it is not known to date whether this approach of choosing a kinetic energy cutoff is sufficient to converge the TNACs. Since the TNACs determine the hopping probabilities, differences in the TNACs between different kinetic energy cutoffs may lead to differences in the dynamics. For all kinetic energy cutoffs, we use a single precomputed nuclear trajectory within the neglect of back reaction approximation (NBRA); thus, any difference in the TNACs is due only to the change in cutoff.

In a TDDFT framework, the first strongly absorbing state of tetracene is a linear combination of multiple SDs.^{48,49} However, since our focus is on understanding how the kinetic energy cutoff changes the TNACs in a molecule, we focus on the TNACs between SDs; the full TDDFT excited states are too computationally expensive to be practical, particularly for the higher kinetic energy cutoffs. We focus first on the TNAC between the (HOMO $- 1 \rightarrow$ LUMO $+ 1$) and (HOMO $- 1 \rightarrow$ LUMO $+ 2$) SDs, which has one of the largest TNACs on average. To visualize the convergence of the TNACs, we use correlation plots between the TNACs calculated by using different kinetic energy cutoffs (Figure 4). For different geometric pairs within this trajectory, the value of this TNAC in our reference data (kinetic energy cutoff of 110 Ry) ranges from 0.08 to 634.98 meV, with a mean of 70.26 meV. At the lowest kinetic energy cutoff of 20 Ry, the average value of this TNAC is 75.27 meV, in reasonable agreement with the reference value. However, there is significant scatter between the TNACs computed at 20 Ry and the reference data, with some geometry pairs having deviations larger than 300 meV. Since this cutoff is far below what is typically used in plane-wave calculations, these large deviations are unsurprising.

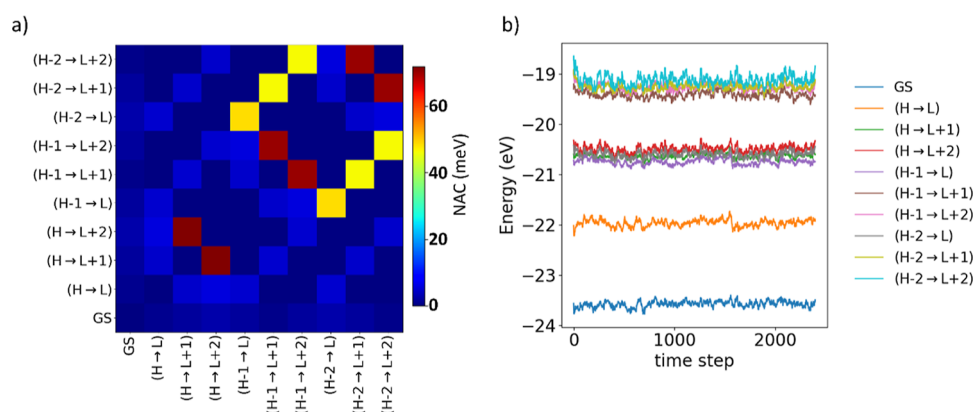


Figure 2. (a) Average TNACs over the geometric trajectory and (b) energy of SDs over time calculated using the PBE functional with a 110 Ry kinetic energy cutoff.

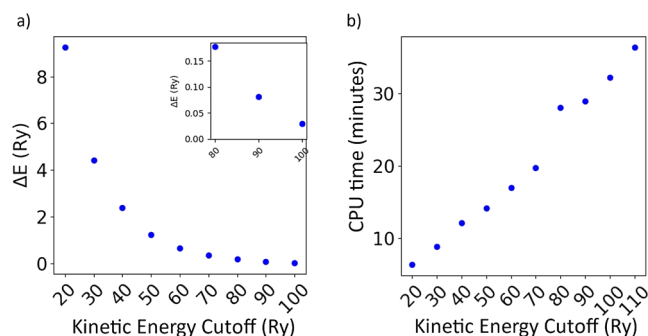


Figure 3. (a) Total energy convergence, zoom in on total energy convergence (inset) and (b) average CPU time for each single-point energy calculation within the 2400 time step trajectory as a function of kinetic energy cutoff.

As the kinetic energy cutoff increases, the agreement between the TNACs and the reference data improves as the points cluster closer to the line of best fit and the number of outliers decreases (Figure 4). We can quantify this agreement using the slope, intercept, and R^2 value for the line of best fit; perfect correlation implies a slope of 1, an intercept of 0, and an R^2 value of 1. In practice, for all SD pairs, all three of these values generally approach their ideal values as the kinetic energy cutoff increases. That improvement is not always monotonic; for example, an increase in cutoff from 50 to 60 Ry (Figure 4d,e) leads to an increase in the intercept and a change in slope from slightly less than 1 to slightly more than 1. We focus on R^2 to determine the level of agreement between the TNACs from a particular cutoff and our reference data as shown in Figure 4j. The red dotted line in Figure 4j is at 0.995, marking a 0.5% error threshold from a perfectly correlated system, where at or above this point we consider the TNAC to be converged. For this particular TNAC, the R^2 value increases sharply as the kinetic energy cutoff increases from 20 to 40 Ry and passes the convergence threshold at 40 Ry; for larger kinetic energy cutoffs, R^2 remains very close to 1.

We now examine the remaining TNACs within our SD basis to examine their convergence with the kinetic energy cutoff. We focus only on the evolution of R^2 with a kinetic energy cutoff; correlation plots for these TNACs are included in Supporting Information (Figures S1–S20). Within the basis of 10 SD, there are a total of 45 unique unordered pairs with different TNACs, with average values ranging from 0.002 to 71 meV (Figure 2a). Since larger TNACs correspond to more

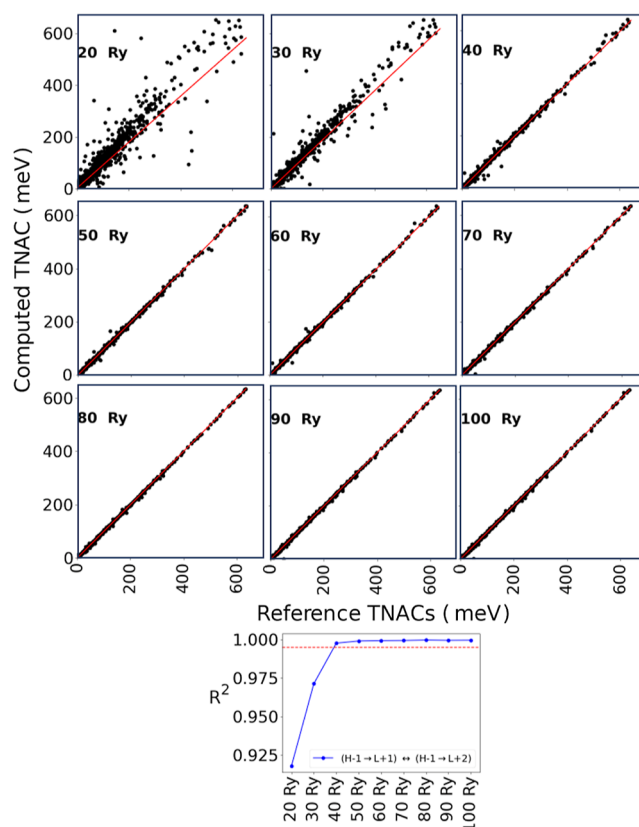


Figure 4. (Top) Correlation plots for TNAC between the (HOMO – 1 → LUMO + 1) and (HOMO – 1 → LUMO + 2) SDs, comparing TNACs at different kinetic energy cutoffs (y axis) to the 110 Ry benchmark data (x axis). (Bottom) R^2 values for the correlation of the TNACs at various kinetic energy cutoffs with the 110 Ry benchmark data.

probable transitions, we focus here on the 20 pairs with TNACs that average at least 2 meV in our reference data. Most of the SD pairs with average TNACs smaller than this cutoff involve transitions in both the occupied and the unoccupied MOs (for example, (HOMO → LUMO) and (HOMO – 1 → LUMO + 1)), so it is unsurprising that their coupling is very small. We consider a TNAC converged if its correlation with the reference data yields $R^2 \geq 0.995$.

We focus first on the TNACs with the largest average magnitude (40–75 meV), shown in Figure 2a, which

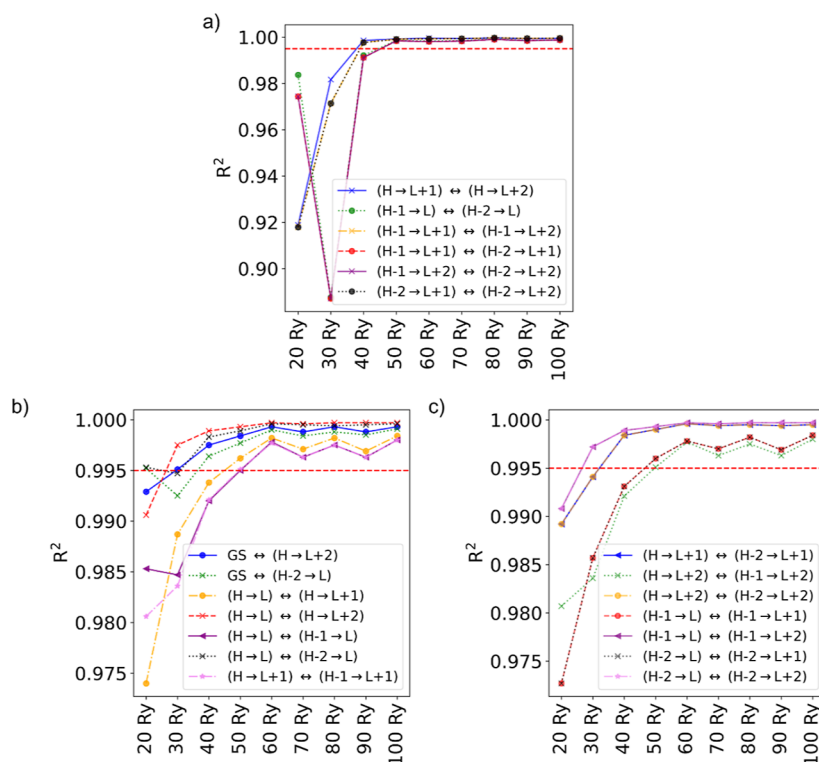


Figure 5. Correlation coefficients from correlation plots of TNACs computed with different kinetic energy cutoffs in the SD basis. (a) TNACs that have a large magnitude (red or yellow squares in Figure 2a). (b,c) TNACs with smaller magnitudes (light blue squares in Figure 2a). All TNACs in (a–c) are ordered by their matrix index from Figure 2a (origin is bottom left corner).

correspond to the red and yellow squares in Figure 2a. All these TNACs are between states that are nearly degenerate because each transition involves a change in MOs within a nearly degenerate pair: either LUMO + 1 \leftrightarrow LUMO + 2 or HOMO – 1 \leftrightarrow HOMO – 2. These TNACs follow two distinct convergence patterns. The three TNACs involving LUMO + 1 \leftrightarrow LUMO + 2 transitions all have R^2 values around 0.92 for the lowest kinetic energy cutoff of 20 Ry, and R^2 increases somewhat at 30 Ry before reaching the convergence threshold of >0.995 at a cutoff of 40 Ry; the SD pair shown in Figure 4 was one of these three TNACs. In contrast, the three TNACs involving HOMO – 1 \leftrightarrow HOMO – 2 transitions have their lowest R^2 values at a 30 Ry cutoff and reach the convergence threshold at a cutoff of 50 Ry. For all six of these TNACs, R^2 is above the convergence threshold at all kinetic energy cutoffs ≥ 50 Ry. Interestingly, the large TNACs have much smaller values of R^2 for the lowest two kinetic energy cutoffs than the smaller TNACs as shown in Figure 5b,c. This suggests that nearly degenerate orbital pairs may be more sensitive to changes in the basis set than nondegenerate orbitals. However, large TNACs also imply that the hopping probability between that pair of states will be large and thus the decay time scales will be fast. We will examine the relationship between the convergence of the TNACs and the convergence of the corresponding decay time scales in the next section.

We now examine the TNACs with smaller average magnitudes (2–6 meV) as shown in Figure 5b,c; this molecule has no TNACs with average values between 6 and 40 meV. The three TNACs that involve a transition between the LUMO and LUMO + 1 all have R^2 values around 0.975 for the lowest kinetic energy cutoff of 20 Ry, and R^2 increases until reaching the convergence threshold at a cutoff of 50 Ry; at larger kinetic energy cutoffs, the R^2 values of these TNACs

oscillate above the convergence threshold. The three TNACs involving a HOMO \leftrightarrow HOMO – 1 transition similarly reach convergence at 50 Ry but oscillate above the convergence threshold at larger kinetic energy cutoffs. In contrast, the three TNACs involving HOMO \leftrightarrow HOMO – 2 transitions converge at 40 Ry, and the three TNACs involving a transition from LUMO \leftrightarrow LUMO + 2 converge at 30 Ry; all of these TNACs have larger and more consistent R^2 values in comparison to the previous sets. A few of these TNACs are above the convergence threshold at 20 Ry but dip below the convergence threshold at 30 Ry. Because of the variation in the cutoff required to reach convergence for different TNACs, an intermediate kinetic energy cutoff may be sufficient to obtain accurate results for some but not all of the TNACs. However, since the convergence threshold we selected is somewhat arbitrary, convergence of the TNACs based on this threshold does not necessarily imply that the relaxation dynamics are converged to the same level. We will examine the relationship between convergence of TNACs and convergence of dynamics in the following section.

3.2. FSSH Dynamics in the SD Basis. After examining the effect of the kinetic energy cutoff on the TNACs, we now turn to the effect of this cutoff on the relaxation times. This will give insights into how tightly converged the TNACs need to be to obtain quantitatively or qualitatively converged relaxation times. To ensure that the variation in our relaxation times is due to differences in the TNACs and not on the stochastic contributions of the surface hopping algorithm, our relaxation times are computed based on five independent FSSH simulations, each using 2500 stochastic realizations of our FSSH simulations at 80 different start times. The standard deviation of the relaxation time out of each initial state is less than 5% of the relaxation time, and in most cases less than 1%,

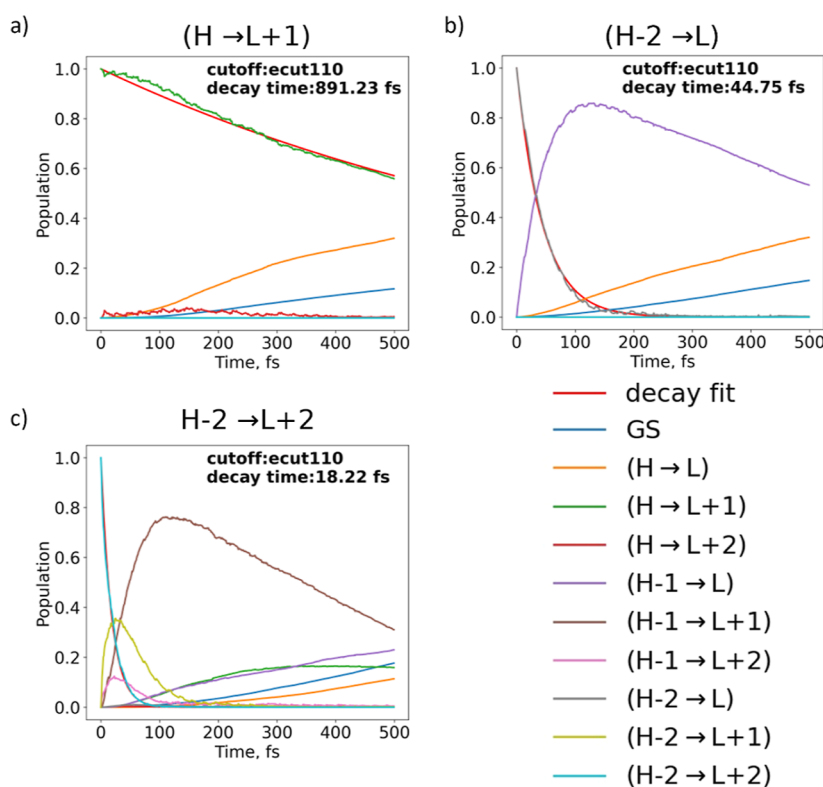


Figure 6. Time evolution of the population for the (a) (HOMO \rightarrow LUMO + 1), (b) (HOMO $-$ 2 \rightarrow LUMO), and (c) (HOMO $-$ 2 \rightarrow LUMO + 2) initial states calculated with 110 Ry kinetic energy cutoff.

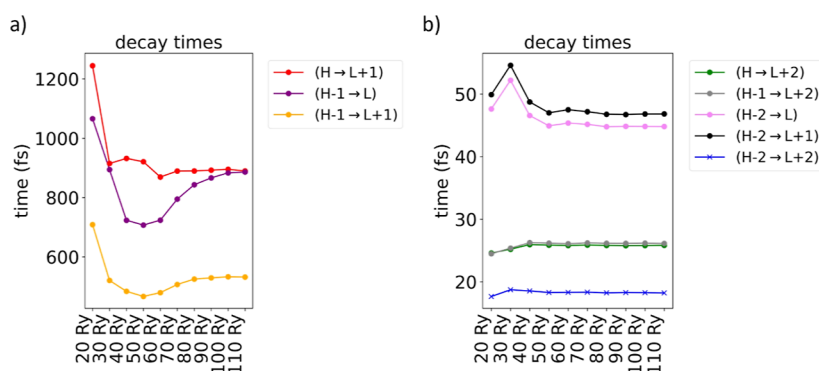


Figure 7. Simulated decay times from five FSSH simulations, with 2500 stochastic realizations, 80 initial starting points, over 500, and 1 fs time steps to obtain error bars for decay times. Error bars are present for all traces.

confirming that the number of trajectories is large enough to produce consistent results with the supplied FSSH simulation parameters. As before, we use the average relaxation time produced from the FSSH simulation using the 110 Ry TNACs as a benchmark.

Several examples of the change in population over time from various initial states are shown in Figure 6. When the initial state is the (HOMO \rightarrow LUMO + 1) SD, the relaxation time is relatively long at \sim 890 fs, and the population transfers primarily to the (HOMO \rightarrow LUMO) and ground-state SDs. This slow decay is unsurprising since (HOMO \rightarrow LUMO + 1) has small TNACs to both the (HOMO \rightarrow LUMO) and ground-state SDs (Figure 2). The (HOMO $-$ 2 \rightarrow LUMO) SD decays much more quickly than the (HOMO $-$ 1 \rightarrow LUMO) SD because the TNAC between those SDs is much larger. Following population transfer to (HOMO $-$ 1 \rightarrow LUMO), the system decays more slowly into (HOMO \rightarrow

LUMO) and the ground state. The (HOMO $-$ 2 \rightarrow LUMO + 2) SD has relatively large TNACs to multiple SDs, resulting in multiple decay pathways involving fast population transfer to the (HOMO $-$ 2 \rightarrow LUMO + 1), (HOMO $-$ 1 \rightarrow LUMO + 2), and (HOMO $-$ 1 \rightarrow LUMO + 1) SDs and slower decay into lower-energy SDs. The decay pathways for the other initial SDs are shown in the Supporting Information.

To evaluate the convergence of the dynamics, we focus solely on the relaxation time out of the initial state; the relative importance of the different decay pathways for each initial state is largely consistent for all kinetic energy cutoffs. We do not consider the (HOMO \rightarrow LUMO) SD as an initial state because the ground state is the only energetically accessible SD for decay, and the TNAC between the ground-state and excited SDs is not well justified for a singly excited method. For all initial states, the absolute value of the % error for all kinetic energy cutoffs is $<40\%$, meaning that all kinetic energy

cutoffs consistently give qualitative agreement with the reference data. This is unsurprising given that the kinetic energy cutoff has a relatively small effect on the average TNAC over the course of the trajectory, even if the TNAC at each particular time step may have significant error. To make consistent comparisons, we consider the relaxation time to be semiquantitatively converged when the absolute value of the % error is <10% and quantitatively converged when the absolute value of the % error is <2%.

Among the higher-energy initial states, the SDs that involve only MOs within an active space from HOMO – 1 to LUMO + 1 have decay times between 400 and 1250 fs (Figure 7a). The three initial SDs have small differences in the convergence pattern of the dynamics with kinetic energy cutoff. The relaxation time of the HOMO → LUMO + 1 SD reaches semiquantitative convergence (<10% error) at 30 Ry, whereas the HOMO – 1 → LUMO and HOMO – 1 → LUMO + 1 SDs do not consistently reach semiquantitative convergence until a kinetic energy cutoff of 60 Ry. All three of these initial SDs reach quantitative convergence (<2% error) of the relaxation times at kinetic energy cutoffs of 60–80 Ry. For these three initial states, the TNACs involved in the main decay pathways all converged at 50 Ry or lower. This suggests that the TNACs must converge to a tighter cutoff than the 0.995 value used in the previous section before quantitative convergence of the decay dynamics is achieved.

For all SDs that involve either HOMO – 2 or LUMO + 2 (Figure 7b), the relaxation times are below 60 fs; the fast decay is consistent with the much larger magnitudes of the TNACs and the near-degeneracy of the MOs. The HOMO – 2 → LUMO + 1 and HOMO – 2 → LUMO SDs achieve semiquantitative convergence at a kinetic energy cutoff of 40 Ry and quantitative convergence at 50 Ry. For both states, the main initial decay pathway involves a transition between HOMO – 2 and HOMO – 1; as shown previously, the TNACs involving this transition have particularly large errors at 30 Ry. In contrast, all three SDs involving LUMO + 2 are semiquantitatively converged for the smallest kinetic energy cutoff of 20 Ry and reach quantitative convergence at 40 Ry. This semiquantitative convergence occurs despite the fact that the TNACs for the relevant decay pathways have much smaller R^2 values at 20 and 30 Ry than most of the smaller TNACs. This suggests that achieving tight convergence of the TNACs may be less important for states that decay quickly than for states with slower decay.

4. CONCLUSIONS

FSSH simulations are a powerful tool to explore the photophysical properties of molecules and materials. Because of the high computational expense of these simulations, many approximations must be made when computing the TNACs. Here, we examined the effect of the kinetic energy cutoff on TNACs and the resulting decay times for a prototypical model system. Higher kinetic energy cutoffs give a more complete description of the orbitals and thus are expected to yield more accurate TNACs.

For our model system, all TNACs in the SD basis converge to an R^2 value > 0.995 by a kinetic energy cutoff of 50 Ry, and many TNACs converge as early as 30–40 Ry. Interestingly, the largest TNACs are the farthest from this convergence threshold at very small kinetic energy cutoffs (20–30 Ry). However, even when R^2 is relatively small, the average TNAC

across the trajectory is in good agreement with the high kinetic energy cutoff reference data.

The choice of kinetic energy cutoff does not significantly change the decay pathway for each initial state. However, the kinetic energy cutoff does have some effect on the time scale of the decay. Initial states with large energy gaps to neighboring lower-energy states and smaller TNACs have slower decay times that depend more strongly on the kinetic energy cutoff. Even for the initial states with the largest variation in decay times, the smallest kinetic energy cutoff of 20 Ry yields decay times within 40% of the reference value. Achieving quantitative convergence (<2% error) of the decay times relative to the reference data requires kinetic energy cutoffs as high as 80 Ry for a few initial states, which is much higher than the cutoff that was required to obtain convergence of the TNACs. In contrast, the initial states that are strongly coupled to lower-energy states showed much less dependence of their decay time scales on the kinetic energy cutoff and reached quantitative convergence at 50 Ry or lower.

These results suggest that the FSSH algorithm using TNACs is overall quite robust to the kinetic energy cutoff. Even at the lowest kinetic energy cutoff of 20 Ry, which is far lower than that used in practice, all of the initial states we examined had decay times within 40% of their converged values. Since the largest errors in decay time scales were seen for the lowest-energy initial states with the fewest possible decay pathways, we expect this result to also hold for higher-energy initial states than those included in our basis set. Given that the choice of functional can change TNACs by up to an order of magnitude, the magnitude of changes with the kinetic energy cutoff is surprisingly small. This suggests that one has a great deal of flexibility in choosing an appropriate kinetic energy cutoff. A larger cutoff is more important for initial states that are energetically isolated or weakly coupled to other states or if one wants results that are quantitatively converged with respect to the kinetic energy cutoff. Since decay time scales between states close in energy are the least sensitive to the kinetic energy cutoff, we expect that as the system size and the density of states increase, the influence of the kinetic energy cutoff will likely decrease. Since decoherence corrections tend to lead to slower decay, further work may be needed to test whether the decay time scales with decoherence corrections are more sensitive to the kinetic energy cutoff. Given the number of other approximations inherent in FSSH simulations, the qualitative convergence that can be achieved with small kinetic energy cutoffs is likely adequate for many applications, and the lower computational cost of a small cutoff facilitates the application of FSSH to a wide variety of chemical systems.

■ ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at <https://pubs.acs.org/doi/10.1021/acs.jpca.3c04858>.

Correlation plots for TNACs at different kinetic energy cutoffs; SD populations over time as a function of kinetic energy cutoff for various initial SDs; correlation coefficients and average TNAC magnitudes from correlation plots; decay times of initial SDs as a function of kinetic energy cutoff; and system temperature vs time from molecular dynamics trajectory (PDF)

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Notes

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