

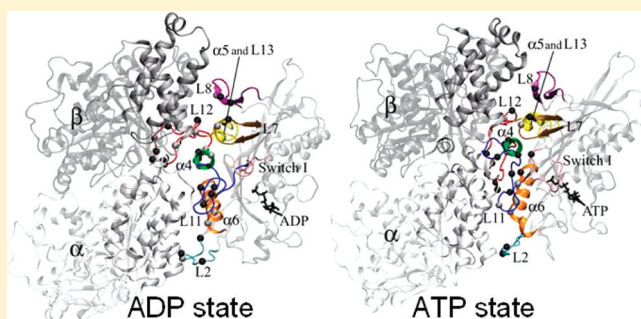
Probing the Structural and Energetic Basis of Kinesin–Microtubule Binding Using Computational Alanine-Scanning Mutagenesis

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S Supporting Information

ABSTRACT: Kinesin–microtubule (MT) binding plays a critical role in facilitating and regulating the motor function of kinesins. To obtain a detailed structural and energetic picture of kinesin–MT binding, we performed large-scale computational alanine-scanning mutagenesis based on long-time molecular dynamics (MD) simulations of the kinesin–MT complex in both ADP and ATP states. First, we built three all-atom kinesin–MT models: human conventional kinesin bound to ADP and mouse KIF1A bound to ADP and ATP. Then, we performed 30 ns MD simulations followed by kinesin–MT binding free energy calculations for both the wild type and mutants obtained after substitution of each charged residue of kinesin with alanine. We found that the kinesin–MT binding free energy is dominated by van der Waals interactions and further enhanced by electrostatic interactions. The calculated mutational changes in kinesin–MT binding free energy are in excellent agreement with results of an experimental alanine-scanning study with a root-mean-square error of ~ 0.32 kcal/mol [Woehlke, G., et al. (1997) *Cell* 90, 207–216]. We identified a set of important charged residues involved in the tuning of kinesin–MT binding, which are clustered on several secondary structural elements of kinesin (including well-studied loops L7, L8, L11, and L12, helices $\alpha 4$, $\alpha 5$, and $\alpha 6$, and less-explored loop L2). In particular, we found several key residues that make different contributions to kinesin–MT binding in ADP and ATP states. The mutations of these residues are predicted to fine-tune the motility of kinesin by modulating the conformational transition between the ADP state and the ATP state of kinesin.



Kinesins, a superfamily of microtubule-based motor proteins, are involved in many important cellular functions, including cell division and the transport of organelles, vesicles, and other cellular cargo.^{2,3} Kinesin converts the chemical energy liberated from ATP hydrolysis to mechanical work that generates movement along the microtubule (MT) track,² which is composed of repeating dimeric subunits formed by an α - and β -tubulin monomer.⁴ There are at least 14 classes of kinesins.^{5,6} A kinesin molecule consists of a catalytic core domain that exhibits MT-stimulated ATPase activity,⁷ a neck region serving as a mechanical amplifier,⁸ a stalk domain for dimerization, and a tail domain for cargo binding.

The primary biochemical pathway of a monomeric kinesin in the presence of MT has been outlined by extensive kinetic studies.^{9–13} Starting from a weak MT-binding ADP state, MT binding to kinesin stimulates ADP release,¹⁴ which leads to a strong MT-binding nucleotide-free state. Then ATP binding is thought to trigger the generation of force via the docking of the neck linker along the catalytic core domain.^{8,15} ATP hydrolysis and subsequent release of inorganic phosphate (P_i) return kinesin to the ADP state. The kinetic parameters of the states and transitions mentioned above vary greatly among different kinesins, which give rise to diverse motor properties. For example, although most kinesins have low MT binding affinities in the ADP state, KIF1A has enhanced affinity for MT in the

ADP state,¹⁶ which underscores its unusual ability to diffuse processively along MT as a monomer.¹⁷

The structural basis of kinesin kinetics remains elusive because detailed structural information is not available for all biochemical states. High-resolution crystal structures were determined only in the absence of MT for kinesins bound to various nucleotide analogues.^{16,18,19} To date, a strongly bound kinesin–MT complex can be modeled only by fitting a kinesin crystal structure and an MT model^{20,21} into cryo-electron microscopy (EM) maps of kinesin-decorated MT filaments.^{15,18,22–26} The EM-based modeling has shed some light on the interface between kinesin and MT, where several secondary structural elements of kinesin are in the proximity of α - and β -tubulin (see Figure 1). The kinesin–MT interface is populated by many charged residues (primarily positive in kinesin and negative in MT), which suggests that electrostatics plays an important role in kinesin–MT binding. However, it remains unknown how much is contributed by electrostatic force and other forces (such as van der Waals interaction), and how much is contributed by individual residues.

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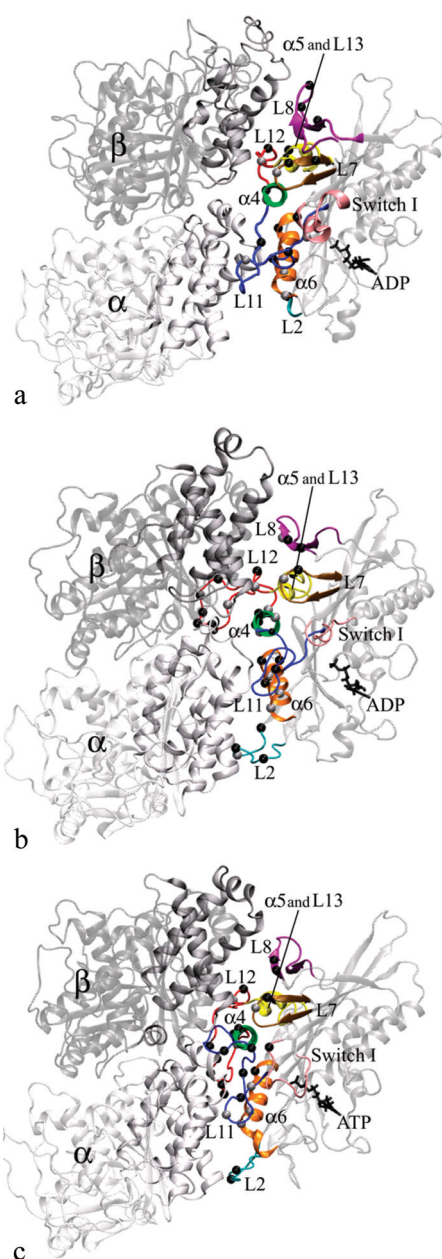


Figure 1. Average structures based on the last 10 ns of MD simulations of (a) KHC_{ADP}, (b) KIF1A_{ADP}, and (c) KIF1A_{ATP}. Kinesin, α -tubulin, and β -tubulin are colored silver, white, and gray, respectively. The key secondary structural elements of kinesin are colored as follows: loop L2 in cyan, loop L7 in ochre, loop L8 in purple, switch I in pink, loop L11 in blue, helix α 4 in green, loop L12 in red, helix α 5 and loop L13 in yellow, and helix α 6 in orange. The charged kinesin residues predicted to be involved in kinesin–MT binding are shown as spheres (positively and negatively charged residues are colored black and silver, respectively). For details of interactions involving these residues, see Figures S1–S3 of the Supporting Information.

Computer modeling techniques, ranging from kinetic modeling^{27,28} to structure-based coarse-grained modeling^{29–33} and all-atom simulations,³⁴ have complemented experimental efforts to probe kinesin motor function. Molecular dynamics (MD) simulations³⁴ based on all-atom force fields³⁵ provide a powerful tool for investigating protein dynamics in the presence of solvents for tens of nanoseconds. Previously, several MD

simulations were performed to investigate the structure and function of various kinesins in the absence of MT.^{36–39} In a recent study, Aprodu et al.⁴⁰ investigated kinesin–MT binding in four nucleotide states (nucleotide-free, ATP, ADP-P_i, and ADP states) by performing 1 ns steered molecular dynamics simulations. However, detailed calculation of the kinesin–MT binding free energy remains to be done, and short MD simulations may be inadequate for probing the dynamic interactions between kinesin and MT.

To probe kinesin–MT binding at the residue level of detail, we have performed computational alanine-scanning mutagenesis to calculate the change in kinesin–MT binding free energy upon mutation of a charged kinesin residue to alanine, which is based on long-time MD simulations of the wild-type kinesin–MT complex in both ADP and ATP states. Computational alanine-scanning mutagenesis has been widely used to dissect the roles of individual residue side chains in protein–protein and protein–ligand binding.^{41–51} First, we have built three kinesin–MT models: human conventional kinesin (KHC) bound to ADP⁵² and mouse KIF1A bound to ADP and ATP.⁵³ KHC and KIF1A were chosen because of the availability of their crystal structures and biochemical data. Then we have performed 30 ns all-atom MD simulations with explicit solvents followed by computational alanine-scanning mutagenesis of all charged residues of the kinesin motor domain. We have found that the kinesin–MT binding free energy is dominated by van der Waals interactions and further enhanced by electrostatic interactions. We have identified an important set of charged residues involved in the tuning of kinesin–MT binding, which are clustered on several secondary structural elements of kinesin (see Figure 1). The calculated mutational changes in kinesin–MT binding free energy are in excellent agreement with results of a previous alanine-scanning mutagenesis study.¹ In particular, we have found several key residues contributing differently to kinesin–MT binding in the ADP and ATP states. The mutations of these residues are predicted to modulate the MT binding affinity of these two states, thereby affecting the motility properties of kinesin.

METHODS

Model Preparation. Three models of the kinesin–MT complex were obtained in previous studies by fitting a kinesin crystal structure into a cryo-EM map of kinesin-decorated MT.^{52,53} The first model (named KHC_{ADP}) consists of an ADP-bound human conventional kinesin [Protein Data Bank (PDB) entry 1BG2] in a complex with α - and β -tubulin.⁵² The second model (named KIF1A_{ADP}) consists of an ADP-bound mouse KIF1A (PDB entry 1I5S) in a complex with α - and β -tubulin.⁵³ The third model (named KIF1A_{ATP}) consists of an ATP analogue-bound mouse KIF1A (PDB entry 1VFV) in a complex with α - and β -tubulin.⁵³ A nucleotide (ADP or ATP) and an Mg²⁺ ion are modeled at the active site. We use the MODLOOP server⁵⁴ to model the missing residues of KHC (residues 1 and 2), KIF1A (residues 1–3, 256–260, and 290–303 of KIF1A_{ADP} and residues 1, 2, 206–212, 254–268, and 289–302 of KIF1A_{ATP}), and MT (residues 1 and 440–451 of α -tubulin and residues 1 and 438–455 of β -tubulin). A long missing region (residues 35–60) in α -tubulin is modeled on the basis of the structure of the corresponding region in β -tubulin. The hydrogen atoms are added with VMD.⁵⁵ All models are immersed into rectangular boxes of water molecules extending up to 10 Å from the proteins in each direction by using VMD. To ensure an ionic concentration of 150 mM and zero net

charge, 70 Na⁺ and 14 Cl[−] ions, 69 Na⁺ and 21 Cl[−] ions, and 64 Na⁺ and 17 Cl[−] ions are added to KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP}, respectively, with VMD. In total, there are 107983, 117503, and 104920 atoms in KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP}, respectively.

MD Simulation Protocol. The three kinesin–MT systems are optimized and equilibrated using energy minimization and MD simulation. First, a 2000-step energy minimization is conducted using the steepest descent method. Then the systems are subjected to a 30 ns unconstrained MD simulation performed in the NPT ensemble. The Nosé–Hoover method⁵⁶ is used with a temperature of 300 K and a pressure of 1 atm. Periodic boundary conditions are applied to the systems. A 10 Å switching distance and a 12 Å cutoff distance are used for nonbonded interactions. The particle mesh Ewald (PME) method⁵⁷ is used to calculate long-range electrostatic interactions. The SHAKE algorithm⁵⁸ is used to constrain bond lengths of hydrogen-containing bonds, which allows a time step of 2 fs for MD simulations. The coordinates of the systems are saved every 1 ps during MD simulations for later analysis. The energy minimization and MD simulation are conducted with NAMD version 2.7b2⁵⁹ using the CHARMM27 force field⁶⁰ and TIP3P water model.⁶¹

Estimation of Experimental Kinesin–MT Binding Free Energy ΔG_{exp} and Its Mutational Change $\Delta\Delta G_{\text{exp}}$. In an experimental alanine-scanning mutagenesis study,¹ 36 residues of KHC were mutated to alanine and the mutants were characterized by ATPase assays that measured the apparent Michaelis–Menten constant for microtubule-dependent ATPase activation (K_{mMT}) and the maximal ATP turnover rate (k_{cat}). On the basis of kinetic analysis,¹ it was argued that $K_{\text{mMT}} \sim k_{\text{cat}}K_{\text{d}}(\text{D})/k_4$, where $K_{\text{d}}(\text{D})$ is the equilibrium dissociation constant for microtubules in the ADP state and k_4 represents the first-order rate constant for ADP release. Thus, using $K_{\text{mMT}} \sim 1.1 \mu\text{M}$, $k_{\text{cat}} \sim 27 \text{ s}^{-1}$,¹ and $k_4 \sim 300 \text{ s}^{-1}$,¹⁰ one can estimate $\Delta G_{\text{exp}} \sim k_{\text{B}}T \ln(K_{\text{mMT}} \times k_4/k_{\text{cat}}) \sim -6.8 \text{ kcal/mol}$. Alternatively, using the measured apparent dissociation constant of KHC construct K332 in the presence of ADP [$K_{\text{d}}(\text{D}) \sim 16 \mu\text{M}$]¹¹, one can estimate $\Delta G_{\text{exp}} = k_{\text{B}}T \ln[K_{\text{d}}(\text{D})] \sim -6.6 \text{ kcal/mol}$.

In ref 1, the ratios of K_{mMT} and k_{cat} between each alanine mutant and the wild type were measured [denoted as K_{mMT}' and k_{cat}' , respectively (see Table 1 of ref 1)]. Thus, the mutational change in binding free energy ($\Delta\Delta G_{\text{exp}}$) for each alanine mutant can be estimated as $k_{\text{B}}T \ln(K_{\text{mMT}}'/k_{\text{cat}}')$.

From the measurement of apparent dissociation constants of KIF1A [$K_{\text{d}} \sim 4.2$ and 6.8 nM in the presence of AMPPNP and ADP, respectively (see ref 16)], one can derive ΔG_{exp} values of approximately -11.6 and -11.3 kcal/mol in the ATP and ADP states, respectively.

Computational Alanine-Scanning Mutagenesis Based on Kinesin–MT Binding Free Energy Calculation. The computational alanine-scanning protocol is a widely used method for evaluating the contributions of individual residue side chains to protein–protein or protein–ligand binding free energy.^{41–51} This protocol is performed as follows. (1) We extract 100 snapshots of the last 10 ns of the MD trajectory of the wild-type kinesin–MT complex (after stripping all waters and ions). (2) For each snapshot, we computationally mutate each charged residue of kinesin to alanine (following ref 42). (3) For each snapshot, we calculate kinesin–MT binding free energy ΔG (see below) for both the wild type (ΔG_{wt}) and each alanine mutant (ΔG_{mut}). The difference in ΔG ($\Delta\Delta G = \Delta G_{\text{mut}} - \Delta G_{\text{wt}}$) is used to evaluate the contribution of each charged

residue to kinesin–MT binding. (4) For each mutant, we calculate the average ($\Delta\Delta G_{\text{avg}}$) and standard error ($\Delta\Delta G_{\text{std}}$) of the $\Delta\Delta G$ values for 100 snapshots and then select those important residues with statistically significant $\Delta\Delta G$ values that satisfy the relationship $|\Delta\Delta G_{\text{avg}}|/\Delta\Delta G_{\text{std}} > 1$.

Following early studies,^{62–65} the same MD trajectory of the wild-type kinesin–MT complex is used to extract snapshots of unbound kinesin, unbound MT, the kinesin–MT complex, and their mutants. Because of a reduction in statistical uncertainty and the cancellation of errors in the ΔG calculation, such a single-trajectory-based protocol was shown to be more efficient and accurate than running separate MD trajectories for bound and unbound proteins and their mutants.^{66–68}

In this study, the kinesin–MT binding free energy is calculated using a continuum solvent model (following ref 69). The binding free energy ΔG is expressed as $\Delta G_{\text{np}} + \Delta G_{\text{elec}}$. Here the nonpolar contribution ΔG_{np} ($=\alpha E_{\text{vdw}}$) is empirically written as a fraction ($\alpha < 1$) of the van der Waals (VDW) interaction energy E_{vdw} between kinesin and MT, and the electrostatic contribution ΔG_{elec} ($=\beta \Delta E_{\text{elec}}$) is empirically written as a fraction ($\beta < 1$) of the change in electrostatic energy ΔE_{elec} from unbound kinesin and MT to the kinesin–MT complex. E_{elec} is calculated using the Poisson–Boltzmann (PB) method.^{70,71} An α of 0.158 and a β of 0.153 are optimized by minimizing the root-mean-square error between calculated $\Delta\Delta G$ and experimental $\Delta\Delta G_{\text{exp}}$ (see above). α is small because the gain in favorable VDW interaction energy between kinesin and MT is largely compensated by the loss of VDW interaction energy between free proteins and water ($\alpha = 0.17$ was obtained in ref 69). β is small because the PB calculation based on a single MD trajectory generally overestimates solvation and interaction energies⁷² of charged residues that are involved in buried salt bridges or hydrogen bonds;^{42,64} these residues are likely to undergo significant structural relaxation upon mutation.⁶² An alternative way to correct such overestimation is to use a large dielectric constant for the protein interior,^{43,69,70,73–77} which may however result in inconsistent treatments of electrostatic contributions from charged and uncharged residues.^{74,75,77} For the PB calculation,^{70,71} a dielectric constant ϵ of 4 is used for the protein interior,^{76,78–80} which accounts for some dielectric response due to structural changes not treated explicitly. An ϵ of 4 can also be obtained on the basis of the measurement of dry proteins and peptide powders.^{78,81} A dielectric constant of 80 is used for the exterior aqueous environment. A probe radius of 1.4 Å is used to define the molecular surface corresponding to the dielectric boundary. The salt concentration is set to 40 mM, corresponding to the buffer condition for kinetic measurements.¹ All the PB calculations are performed using the PBEQ module^{71,82,83} of CHARMM.⁸⁴ Each PB calculation is conducted by using the bilinear interpolation to construct the boundary potential. The atomic Born radii used here were previously calibrated and optimized to reproduce the electrostatic free energy of the 20 amino acids in MD simulations with explicit water molecules.⁸³ Following previous studies,^{41–43,62,67,85} the entropic contribution to ΔG is ignored because our goal is to estimate the relative binding free energy ($\Delta\Delta G$) for structurally similar complexes (i.e., wild type and mutants) whose entropy contribution largely cancels out.

Optimization and Evaluation of the Calculated Mutational Change in Binding Free Energy $\Delta\Delta G$. We scan α and β values within the range of 0–1 to minimize the following root-mean-square error (rmse) between the calculated $\Delta\Delta G$

and experimental $\Delta\Delta G_{\text{exp}}$: $\text{rmse} = [1/N \sum_n (\Delta\Delta G_n - \Delta\Delta G_{\text{exp},n})^2]^{1/2}$, where $\Delta\Delta G_n$ ($\Delta\Delta G_{\text{exp},n}$) is the calculated $\Delta\Delta G$ (experimental $\Delta\Delta G_{\text{exp}}$) of residue n and $N = 20$ is the number of charged residues of KHC with $\Delta\Delta G_{\text{exp}}$ measured.¹

The overall fitting quality is assessed by the coefficient of multiple determination R^2 (see ref 86): $R^2 = 1 - [\sum_n (\Delta\Delta G_n - \Delta\Delta G_{\text{exp},n})^2] / [\sum_n (\Delta\Delta G_{\text{exp},n} - \Delta\Delta G_{\text{exp,avg}})^2]$, where $\Delta\Delta G_{\text{exp,avg}}$ is the average of the experimental $\Delta\Delta G_{\text{exp}}$.

To assess possible overfitting, the cross-validated statistical figure of merit, Q_{loo}^2 , is calculated as follows: $Q_{\text{loo}}^2 = 1 - [\sum_n (\Delta\Delta G_n^* - \Delta\Delta G_{\text{exp},n})^2] / [\sum_n (\Delta\Delta G_{\text{exp},n} - \Delta\Delta G_{\text{exp,avg}})^2]$, where $\Delta\Delta G_n^*$ is calculated using the optimal parameters for the data set with $\Delta\Delta G_{\text{exp},n}$ excluded, so Q_{loo}^2 gives a measure of the model's predictive power.

Residue Conservation Analysis. We use precalculated residue conservation scores from ConSurf-DB (<http://consurfd.b.tau.ac.il/>) to assess residue conservation in the kinesin superfamily. We use the ConSurf server (<http://consurf.tau.ac.il/>)^{87–89} to align multiple sequences of kinesin 1 and kinesin 3 families and assess residue conservation in each family. The kinesin sequences are obtained from the Kinesin Homepage (<http://www.cellbio.duke.edu/kinesin/MotorSeqTable.html>). The residue positions with a conservation grade of ≥ 7 are deemed to be conserved.

RESULTS AND DISCUSSION

Equilibration of the MD Simulation. We have built three all-atom models of the kinesin–MT complex [KHC_{ADP} , $\text{KIF1A}_{\text{ADP}}$, and $\text{KIF1A}_{\text{ATP}}$ (see Methods)] comprised of a kinesin (KHC or KIF1A) and two tubulins (α - and β -tubulin), which are optimized by energy minimization followed by 30 ns MD simulations with explicit solvents (see Methods). The root-mean-square deviation (rmsd) of all backbone atoms (with respect to the initial models) slowly increases during the simulations, suggesting that the entire kinesin–MT complex undergoes slow structural relaxation (see Figure 2). Never-

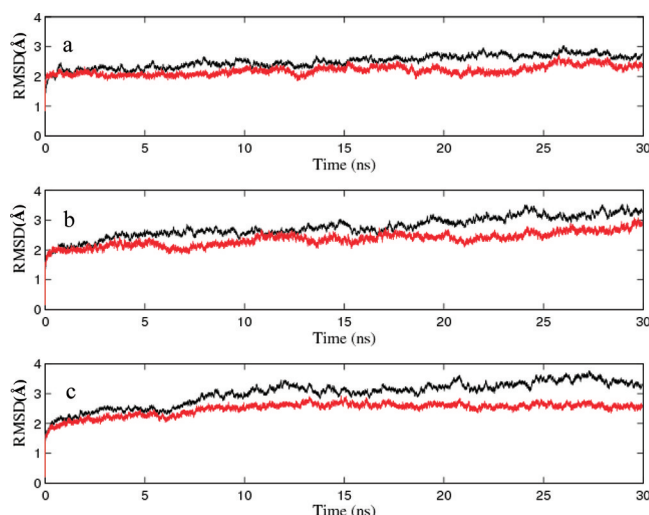


Figure 2. Backbone rmsd (with respect to the initial models) of all residues (black) and kinesin–MT interface residues (red) in (a) KHC_{ADP} , (b) $\text{KIF1A}_{\text{ADP}}$, and (c) $\text{KIF1A}_{\text{ATP}}$ as a function of simulation time.

theless, the rmsd restricted to the kinesin–MT interface (as defined in ref 90) is smaller and more stable during most of the simulations (see Figure 2), so we postulate that the kinesin–

MT dynamic interaction has reached an equilibrium state during the last 10 ns of MD simulations, which allows the calculation of the kinesin–MT binding free energy (see below).

Calculation of the Kinesin–MT Binding Free Energy.

We initially used the MM-PBSA (molecular mechanics–Poisson–Boltzmann surface area) method^{62,91} to calculate the kinesin–MT binding free energy (ΔG) and its change caused by alanine mutation ($\Delta\Delta G$), which gave significantly overestimated $\Delta\Delta G$ values compared with experimental ones. This overestimation is probably due to inadequate structural relaxation in the MM-PBSA calculation based on a single MD trajectory.⁶² To fully account for structural relaxation, one can run separate MD simulations for unbound and bound proteins and their mutants, which will not only increase the computing cost dramatically but also introduce large statistical error into the free energy calculation. To allow accurate and efficient calculation of the kinesin–MT binding free energy, we have adopted an empirical method that calculates ΔG as a linear combination of two contributions: the kinesin–MT VDW interaction energy (calculated by MM force field) and the change in electrostatic free energy upon binding [calculated by the PB method (see Methods)]. The corresponding coefficients [$\alpha = 0.158$ (for the VDW contribution) and $\beta = 0.153$ (for the electrostatic contribution)] are determined by fitting the experimental $\Delta\Delta G_{\text{exp}}$ (see Methods). This empirical method is akin to the linear interaction energy method (LIE),^{92,93} which calculates protein–ligand binding free energy by linearly combining electrostatic and VDW energies between the ligand and its surroundings. Similar to this study, several LIE-based studies also yielded small β values of ≤ 0.2 for the electrostatic contribution.^{94–96} Similarly, in an early PB-based calculation of $\Delta\Delta G$ of 10 lysozyme mutants, it was found that the electrostatic contribution should be reduced by 0.4 to fit the experimental $\Delta\Delta G$ data.⁹⁷ The idea of linearly combining various terms of MM-PBSA was also explored recently.⁹⁸

To calculate kinesin–MT binding free energy ΔG in KHC_{ADP} and $\text{KIF1A}_{\text{ATP}}$, we used 100 snapshots from the last 10 ns of MD trajectories (for $\text{KIF1A}_{\text{ADP}}$, only 80 snapshots are used because ΔG shows drift during the last 2 ns of MD simulation). To check the stability of ΔG values over time (10 ns for KHC_{ADP} and $\text{KIF1A}_{\text{ATP}}$ and 8 ns for $\text{KIF1A}_{\text{ADP}}$), we performed linear regression that yields negligible slopes: 4.3×10^{-4} , 2.8×10^{-4} , and 6.4×10^{-4} kcal mol^{−1} ps^{−1} for KHC_{ADP} , $\text{KIF1A}_{\text{ADP}}$, and $\text{KIF1A}_{\text{ATP}}$, respectively, which are comparable with low values (2×10^{-3} kcal mol^{−1} ps^{−1}) reported in a previous study.⁴⁵

We have calculated the average ΔG (ΔG_{avg}) for three kinesin–MT systems (approximately -23.2 , -30.0 , and -29.4 kcal/mol for KHC_{ADP} , $\text{KIF1A}_{\text{ADP}}$, and $\text{KIF1A}_{\text{ATP}}$, respectively). Estimation based on experimental data gives ΔG_{exp} values of approximately -6.8 , -11.3 , and -11.6 kcal/mol for the ADP state of KHC and the ADP and ATP states of KIF1A, respectively (see Methods). The shift of 16–19 kcal/mol between calculated and experimental ΔG values can be attributed to entropy loss upon kinesin–MT binding, which is not included in our calculation [the rotational and translational entropy loss upon binding was estimated to be 7–15 kcal/mol (see ref 99)]. The calculation described above explains the experimental finding of a higher MT binding affinity for KIF1A than for KHC in the ADP state, which underlies the unique ability of KIF1A to undergo processive diffusion along MT for many steps without dissociation.¹⁷ We note that our MD simulation is too short to explore the

diffusive motion of KIF1A along MT, and it can only probe the affinity of KIF1A for MT while it binds transiently with a tubulin dimer. It is possible that a different affinity would be obtained if diffusion could be properly modeled.

We have performed a breakdown of ΔG to electrostatic (−4.0, −7.9, and −9.0 kcal/mol for KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP}, respectively) and VDW (−19.2, −22.1, and −20.4 kcal/mol for KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP}, respectively) contributions. The kinesin–MT binding free energy is found to be dominated by the VDW contribution (including the nonpolar part of the solvation free energy). The electrostatic contribution (including the polar part of the solvation free energy), although relatively small, can still play an important role in kinesin–MT binding. In fact, the electrostatic contribution is more important than the VDW contribution in determining the difference in ΔG between two kinesins—the electrostatic (VDW) contribution to the difference in ΔG between KHC_{ADP} and KIF1A_{ADP} is 3.9 (2.9) kcal/mol. The small electrostatic contribution is due to the cancellation of favorable electrostatic interactions between oppositely charged residues of kinesin and MT by the unfavorable solvation energy of charged residues.^{100–102} Although the net contribution of electrostatic interactions to the protein assembly process is generally expected to be unfavorable, favorable electrostatic binding free energy may arise when the partner proteins have a complementary charge distribution (see refs 72 and 90), which may result from evolutionary optimization of binding affinity.¹⁰³

Comparison with Experimental Alanine-Scanning Mutagenesis Results. To further probe the kinesin–MT interactions at the residue level of detail, we performed computational alanine-scanning mutagenesis on all charged residues of kinesin (see Methods). The contribution of each charged residue's side chain to kinesin–MT binding is quantified by calculating the change in kinesin–MT binding free energy ($\Delta\Delta G$) caused by its mutation to alanine (see Methods). Furthermore, we have performed a breakdown of $\Delta\Delta G$ into VDW and electrostatic contributions for each charged residue of kinesin, which suggests a larger contribution of electrostatic than VDW interaction for these residues (see Table S1 of the Supporting Information).

In an experimental alanine-scanning mutagenesis study, 36 alanine mutants of KHC were characterized using ATPase assays.¹ Experimental $\Delta\Delta G_{\text{exp}}$ values for alanine mutations can be estimated from the measurements of the apparent Michaelis–Menten constant for microtubule-dependent ATPase activation (K_{mMT}) and the maximal ATP turnover rate (k_{cat}) (see Methods). We have found that the calculated $\Delta\Delta G$ values are in excellent agreement with experimental $\Delta\Delta G_{\text{exp}}$ values with an rmse of 0.35 kcal/mol and an R^2 of 0.72 (see Figure 3). To check for possible overfitting, we redo the optimization of α and β after leaving one $\Delta\Delta G_{\text{exp}}$ value (for residue n) out, and then we recalculate $\Delta\Delta G$ for residue n . In this way, we have obtained a cross-validated statistical figure of merit ($Q_{\text{loo}}^2 = 0.67$) and a leave-one-out cross-validated rmse of 0.38 kcal/mol, which are close to the R^2 and rmse obtained by fitting all $\Delta\Delta G_{\text{exp}}$ data. Therefore, our $\Delta\Delta G$ calculation based on optimized parameters ($\alpha = 0.158$, and $\beta = 0.153$) has good predictivity and gives a statistically robust result.

To further test the predictivity of our $\Delta\Delta G$ calculation, we calculated $\Delta\Delta G$ for 16 uncharged residues of KHC probed in ref 1 (using the model parameters optimized for charged kinesin residues). Among these uncharged residues, we have

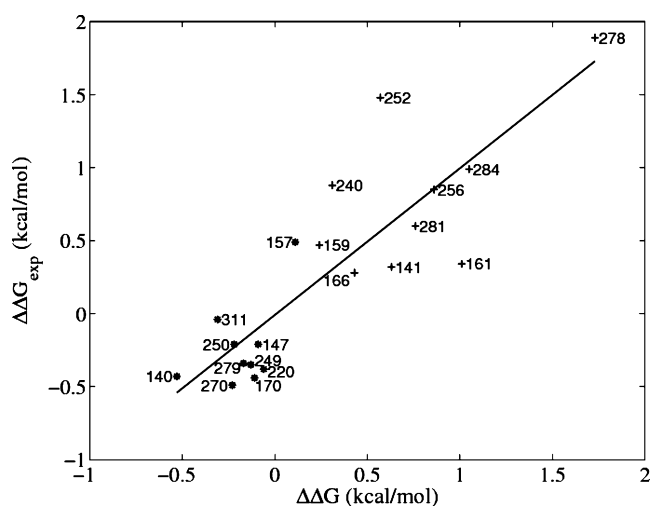


Figure 3. Calculated $\Delta\Delta G$ values and experimental $\Delta\Delta G_{\text{exp}}$ values for 20 charged residues of KHC.¹ Plus signs denote data for positively charged residues, and asterisks denote data for negatively charged residues. The straight line shows the result of linear fitting: $\Delta\Delta G_{\text{exp}} = -0.009 + 1.004\Delta\Delta G$.

identified Y274 and L248 as the top two most important for kinesin–MT binding (ranked by $\Delta\Delta G$, which is 1.24 kcal/mol for Y274 and 0.92 kcal/mol for L248). Encouragingly, the same two residues were identified by experimental alanine-scanning mutagenesis¹ as the top two most important for kinesin–MT binding (ranked by $\Delta\Delta G_{\text{exp}}$, which is 0.87 kcal/mol for Y274 and 0.86 kcal/mol for L248). Additionally, R278 is identified as the most important charged residue for kinesin–MT binding both computationally and experimentally (see Table 1). Therefore, our $\Delta\Delta G$ calculation can correctly identify both charged and uncharged residues critically involved in kinesin–MT binding.

Next, we use $\Delta\Delta G$ values to sort all charged residues of kinesin and then select those residues with large positive or negative $\Delta\Delta G$ values (see Table 1), which are predicted to be involved in strengthening or weakening kinesin–MT binding, respectively. We will focus our discussion on these residues.

Positively Charged Residues Involved in Strengthening Kinesin–MT Binding. A number of positively charged residues are predicted to be important in strengthening kinesin–MT binding, which are distributed in several secondary structural elements of kinesin (see Figure 1), including loop L2 (K41, K44, and K48 of KIF1A), loop L7 (K141 and R143 of KHC and R153 of KIF1A), loop L8 (K159, R161, and K166 of KHC and K163, R167, and R169 of KIF1A), switch I (R203 of KHC and R216 of KIF1A), loop L11 (K237, K240, and K252 of KHC and R254, K261, R264, and K266 of KIF1A), helix $\alpha 4$ (K256 of KHC and K273 and K280 of KIF1A), loop L12 (R278 of KHC and K294, K296, K297, K298, K299, K300, and R307 of KIF1A), helix $\alpha 5$ and loop L13 (K281 and R284 of KHC and R316 of KIF1A), and helix $\alpha 6$ (K313 and R321 of KHC and R346 and R350 of KIF1A).

To offer structural insights into the kinesin–MT interactions involving the charged residues mentioned above, we calculated an average structure for each system from the last 10 ns of the MD trajectory. On the basis of the average structure, we define that two charged residues form electrostatic interaction if their charged side chain atoms (NZ of Lys, NH1 and NH2 of Arg, OE1 and OE2 of Glu, and OD1 and OD2 of Asp) are within 10

Table 1. Mutational Changes in Kinesin–MT Binding Free Energy ($\Delta\Delta G_{\text{avg}}$) of Selected Charged Residues in KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP}^a

KHC _{ADP}		KIF1A _{ADP}		KIF1A _{ATP}		KIF1A _{ADP} – KIF1A _{ATP}	
residue number	$\Delta\Delta G_{\text{avg}}$ (kcal/mol)	residue number	$\Delta\Delta G_{\text{avg}}$ (kcal/mol)	residue number	$\Delta\Delta G_{\text{avg}}$ (kcal/mol)	residue number	$\Delta\Delta G_{\text{avg}}$ (kcal/mol)
R278*	1.73 [1.89]	R346*	1.93	K261	1.85	K44	1.42
K237*	1.41	K44	1.73	K297	1.49	R346	1.25
R284	1.05 [0.99]	K48	1.31	R169*	1.11	K48	0.71
R161	1.01 [0.34]	R307* (278)	1.24	R307* (278)	1.01	K296	0.56
R321*	0.99	K296	1.18	K280*	0.95	D256	0.38
K313	0.92	R350* (321)	1.09	E267* (250)	0.86	R350	0.31
K256*	0.86 [0.85]	R169*	0.99	K273 (256)	0.82	K300	0.31
K281*	0.76 [0.60]	K266	0.96	R350* (321)	0.78	K294	0.30
K141	0.63 [0.32]	K261	0.93	R167	0.75		
K252	0.57 [1.48]	K273 (256)	0.82	K41	0.74		
K166	0.43 [0.28]	K280*	0.66	K266	0.72	K297	–1.25
K44	0.42	K300	0.64	R153 (141)	0.69	E267	–1.06
R203*	0.32	R167	0.63	R346*	0.69	K261	–0.92
K240*	0.31 [0.88]	K298	0.62	K298	0.66	R254	–0.53
K159	0.24 [0.47]	K41	0.58	R254* (237)	0.65	R153	–0.35
R143	0.21	R316*	0.46	K296	0.61	R216	–0.30
		K294	0.45	K48	0.59	K280	–0.29
D140*	–0.53 [–0.43]	K299	0.44	R216* (203)	0.47		
E309	–0.34	R153 (141)	0.34	R316*	0.34		
E311*	–0.31 [–0.04]	R264*	0.31	K300	0.33		
E270	–0.23 [–0.49]	K297	0.24	K44	0.32		
E199	–0.23			R264*	0.27		
E250*	–0.22 [–0.21]	D339	–0.60	K299	0.25		
E236*	–0.20	E170* (157)	–0.40	R18	0.23		
		E45	–0.39	K163 (K150)	0.19		
		D349	–0.39				
		E152* (140)	–0.31	D339	–0.69		
		E340* (311)	–0.27	D256	–0.59		
		D308* (279)	–0.25	E340* (311)	–0.45		
		E287 (270)	–0.24	E152* (140)	–0.41		
		D256	–0.21	E253* (236)	–0.38		
		E267* (250)	–0.20	E170* (157)	–0.32		
		E179 (170)	–0.19	D308* (279)	–0.27		
				E179 (170)	–0.21		

^aWe select charged residues of kinesin for which $|\Delta\Delta G_{\text{avg}}|/\Delta\Delta G_{\text{std}} > 1$, where $\Delta\Delta G_{\text{std}}$ is 0.19–0.20 for KHC_{ADP}, 0.18–0.19 for KIF1A_{ADP}, and 0.19–0.20 for KIF1A_{ATP} and 0.26–0.27 for the difference between KIF1A_{ADP} and KIF1A_{ATP}. The experimental $\Delta\Delta G_{\text{exp}}$ values are enclosed in brackets. The KHC residue numbers corresponding to some KIF1A residues are enclosed in parentheses. Residues marked with an asterisk are conserved in the kinesin superfamily. Residues marked with a caret are conserved in the kinesin 1 or kinesin 3 family.

Å of each other, and two oppositely charged residues form a salt bridge interaction if their charged side chain atoms are within 4 Å of each other (see ref 104). Most of the charged residues mentioned above form favorable electrostatic interactions with MT residues in the average structures. For example, K41 of loop L2 interacts with E155, E196, and R156 of α -tubulin in KIF1A_{ADP} (Figure S2a of the Supporting Information) and with E196, D424, and R264 of α -tubulin in KIF1A_{ATP} (Figure S3a of the Supporting Information). K141 of loop L7 interacts with E159 and E196 of β -tubulin in KHC_{ADP} (Figure S1a of the Supporting Information). R161 of loop L8 interacts with E420, E422, D427, and R400 of β -tubulin in KHC_{ADP} (Figure S1b of the Supporting Information). K237 of loop L11 interacts with E414, E417, and K112 of α -tubulin in KHC_{ADP} (Figure S1c of the Supporting Information). K256 of helix α 4 interacts with E196, R158, and R264 of β -tubulin in KHC_{ADP} (Figure S1d of the Supporting Information). R278 of loop L12 interacts with E196, E420, E422, D427, E431, and R264 of β -tubulin in

KHC_{ADP} (Figure S1e of the Supporting Information). K281 of helix α 5 and loop L13 interacts with E159, E196, and E420 of β -tubulin in KHC_{ADP} (Figure S1f of the Supporting Information). K313 of helix α 6 interacts with E420 and E423 of α -tubulin in KHC_{ADP} (Figure S1g of the Supporting Information).

Interestingly, some of these charged residues interact differently with MT in KIF1A_{ADP} and KIF1A_{ATP}.

In loop L2, K44 interacts with E196, E417, E420, E423, and D424 of α -tubulin to form 10 opposite-charge atom pairs (within 10 Å, same as below) in KIF1A_{ADP} (Figure S2a of the Supporting Information) and with D431, E434, and K430 of α -tubulin to form only four opposite-charge atom pairs and one same-charge pair in KIF1A_{ATP} (Figure S3a of the Supporting Information). Therefore, K44 of loop L2 favors MT binding in the ADP state over the ATP state [with a $\Delta\Delta G$ difference of 1.42 kcal/mol (see Table 1)]. Indeed, previous EM studies showed the proximity of loop L2 to MT in both KIF1A⁵³ and

NCD²² in the ADP state. In contrast, loop L2 of KHC is very short and thus contributes little to kinesin–MT binding.

In loop L8, R167 is involved in MT binding in both ADP and ATP states [with a $\Delta\Delta G$ difference of 0.12 kcal/mol (see Table 1)]; it interacts with E196, E417, and E420 of β -tubulin in KIF1A_{ADP} (Figure S2b of the Supporting Information) and with D414, E417, and E420 of β -tubulin in KIF1A_{ATP} (Figure S3b of the Supporting Information). A nearby residue, R169, also binds MT strongly in the ATP and ADP states [with a $\Delta\Delta G$ difference of 0.12 kcal/mol (see Table 1)]. Therefore, loop L8 is involved in MT binding in ADP and ATP states, which agrees with a mutational study of L8 residues (K161A/R167A/R169A/K183A).¹⁶

In the switch I region (near loop L9), R216 interacts with E414 of α -tubulin in KIF1A_{ATP} (Figure S3c of the Supporting Information) but forms no electrostatic interaction with MT in KIF1A_{ADP}. This interaction in the ATP state may allow MT binding to trigger displacement of switch I toward MT that causes the closing of the active site as found by EPR spectroscopy.^{38,105} Interestingly, the R598A mutant of Kar3 (corresponding to R216 of KIF1A) was found to exhibit no MT-activated ATPase activity in steady state assays and bind weakly to MT in the presence of ADP or in the absence of nucleotide.¹⁰⁶

In loop L11, R254 interacts with E155, E414, E417, and K112 of α -tubulin in KIF1A_{ATP} (Figure S3d of the Supporting Information) and forms no electrostatic interaction with MT in KIF1A_{ADP}. Similarly, K261 interacts favorably with E414 and E417 of α -tubulin in KIF1A_{ATP} (Figure S3d of the Supporting Information) and with E97 of α -tubulin and D163 of β -tubulin, but unfavorably with R105 of α -tubulin in KIF1A_{ADP} (Figure S2c of the Supporting Information). Therefore, R254 and K261 favor MT binding in the ATP state over the ADP state [with $\Delta\Delta G$ difference values of 0.53 and 0.92 kcal/mol (see Table 1)], suggesting that loop L11 is involved in MT binding in the ATP state more than in the ADP state, which agrees with the previous finding that mutations of L11 residues (K261A/R264A/K266A) affect affinity in the ATP state more than in the ADP state.¹⁶

In loop L12 (including K loop of KIF1A), K296 interacts with D437, D441, E442, and E445 of β -tubulin to form seven opposite-charge atom pairs in KIF1A_{ADP} (Figure S2d of the Supporting Information) and with D396, E420, E423, and R422 of α -tubulin to form five opposite-charge atom pairs and two same-charge atom pairs in KIF1A_{ATP} (Figure S3e of the Supporting Information). Therefore, K296 favors MT binding in the ADP state over the ATP state [with a $\Delta\Delta G$ difference of 0.56 kcal/mol (see Table 1)]. In the average structures of the KIF1A–MT complex, loop L12 and K loop extend farther into MT in the ADP state than in the ATP state (see Figure 1b,c), which agrees with the previous finding that mutations of K loop lysines affect affinity in the ADP state more than in the ATP state.¹⁰⁷

In helix $\alpha 6$, R346 interacts with E415, E420, E423, D424, and R402 of α -tubulin in KIF1A_{ADP} (Figure S2e of the Supporting Information) and with only E420 and E423 of α -tubulin in KIF1A_{ATP} (Figure S3f of the Supporting Information). R350 interacts with E414, E415, E420, and R402 of α -tubulin in KIF1A_{ATP} (Figure S3f of the Supporting Information) and with E415, K401, and R402 of α -tubulin in KIF1A_{ADP} (Figure S2e of the Supporting Information). R346 and R350 favor MT binding in the ADP state versus the ATP state [with $\Delta\Delta G$ difference values of 1.25 and 0.31 kcal/mol,

respectively (see Table 1)], so helix $\alpha 6$ is likely involved in MT binding in the ADP state more than in the ATP state.

Taken together, the structural observations described above agree with our finding of different $\Delta\Delta G$ values for the residues mentioned above in KIF1A_{ATP} and KIF1A_{ADP} (see Table 1). These residues are therefore predicted to critically regulate the different MT binding affinity between the ADP and ATP states.

The importance of the charged residues mentioned above is supported by residue conservation analysis based on multiple-sequence alignment (see Methods). Among these residues, K237, K240, K256, K278, K281, and R321 of KHC and R169, R216, R254, R264, K280, R307, R316, R346, and R350 of KIF1A are conserved in the kinesin superfamily. K159, K252, and R284 of KHC are conserved in the kinesin 1 family. K266, K273, K294, and K296 of KIF1A are conserved in the kinesin 3 family.

Negatively Charged Residues Involved in Weakening of Kinesin–MT Binding. A number of negatively charged residues are predicted to be involved in weakening kinesin–MT binding, which are distributed in several key structural motifs of kinesin (see Figure 1), including loop L2 (E45 of KIF1A), loop L7 (D140 of KHC and E152 of KIF1A), loop L8 (E170 and E179 of KIF1A), switch I (E199 of KHC), loop L11 (E236 and E250 of KHC and E253, D256, and E267 of KIF1A), helix $\alpha 4$ (E270 of KHC and E287 of KIF1A), loop L12 (D308 of KIF1A), helix $\alpha 6$ (E309 and E311 of KHC and D339, E340, and D349 of KIF1A).

Most of these residues form unfavorable electrostatic interactions with MT residues in the average structures. For example, D140 of loop L7 interacts with E159, E196, and R158 of β -tubulin in KHC_{ADP} (Figure S1a of the Supporting Information); E236 of loop L11 interacts with E414 of α -tubulin in KHC_{ADP} (Figure S1c of the Supporting Information), and E309 of helix $\alpha 6$ interacts with E420 of α -tubulin in KHC_{ADP} (Figure S1g of the Supporting Information).

Interestingly, some of these charged residues interact differently with MT in KIF1A_{ADP} and KIF1A_{ATP}. For example, in loop L11, E267 interacts with E97, D98, E113, E411, K96, and R105 of α -tubulin and D163, R164, and R253 of β -tubulin in KIF1A_{ATP} (Figure S3g of the Supporting Information) but forms no electrostatic interaction with MT in KIF1A_{ADP}. Therefore, E267 favors kinesin–MT binding in the ATP state over the ADP state [with a $\Delta\Delta G$ difference of 1.06 kcal/mol (see Table 1)], which further contributes to the high affinity in the ATP state together with those positively charged residues of loop L11.

The importance of these charged residues is supported by residue conservation analysis (see Methods). Among these residues, D140, E311, E236, E250 of KHC, E152, E170, E253, E267, D308, and E340 of KIF1A are conserved in the kinesin superfamily.

The importance of the charged residues mentioned above is also supported by previous mutational studies. The E164A mutant of *Drosophila melanogaster* kinesin (corresponding to E170 of KIF1A) was found to exhibit reduced steady state ATPase activity and higher affinity for MT.¹⁰⁸ The E236A mutant of KHC (corresponding to E253 of KIF1A) was shown to have very low ATPase activities in the presence of MT.¹⁰⁹ The mutation of E237 of rat KHC (corresponding to E253 of KIF1A) reduced or abolished MT-dependent ATPase activity.¹¹⁰ The E631A mutant of Kar3 (corresponding to E253 of KIF1A) was shown to exhibit no MT-activated ATPase activity in steady state assays and bind tightly to MT in the

presence of ADP or in the absence of nucleotide.¹⁰⁶ The E311A mutant of KHC was found to show a ~3-fold reduction in both MT gliding velocity and MT-stimulated ATPase activity.¹¹¹

Interactions between Kinesin and the E-Hook of MT.

Finally, we discuss the kinesin–MT interactions involving the C-terminal E-hook of tubulins (residues 440–451 of α -tubulin and 438–455 of β -tubulin). In the average structure of KIF1A_{ADP}, E45 of loop L2 interacts with E443 of α -tubulin (see Figure S2a of the Supporting Information) and D289, K296, and K299 of loop L12 interact with D441, E442, and E445 of β -tubulin (see Figure S2d of the Supporting Information). No electrostatic interactions are formed between the E-hook and kinesin in the other two systems. Our finding supports the proposal that the interaction between loop L12 and the E-hook allows diffusive binding of KIF1A in the ADP state,¹⁶ and this interaction does not contribute to the strong MT-binding state like the ATP state.¹⁰⁷

CONCLUSION

To develop a detailed understanding of kinesin–MT binding in structural and energetic terms, we have performed an extensive alanine-scanning mutagenesis study of all charged residues of three kinesin models (ADP-bound KHC and ADP-bound and ATP-bound KIF1A). This study is based on three 30 ns MD simulations of kinesin–MT systems in ADP and ATP states. To the best of our knowledge, these simulations are the longest ever performed for the kinesin–MT complex with explicit solvents. Although the kinesin kinetics (approximately milliseconds) remains beyond the reach of all-atom MD simulations, far less simulation time may suffice for free energy calculation based on the sampling of conformational fluctuations near a stable state (such as the ADP state and the ATP state).

On the basis of the MD simulations, we have calculated the kinesin–MT binding free energy (ΔG) and its mutational changes ($\Delta\Delta G$). The latter are in excellent agreement with results of an experimental alanine-scanning study.¹ Furthermore, on the basis of the ranking of $\Delta\Delta G$ values, we have predicted a key set of positively and negatively charged residues of kinesin involved in strengthening and weakening kinesin–MT binding, respectively. These charged residues form intricate networks of hydrogen bonds and salt bridges at the kinesin–MT interface, which make an overall favorable contribution to kinesin–MT binding.⁹⁰ Many of the predicted key residues are strongly conserved or previously found to be important in mutational studies.^{1,15,16,106,107,110–112} Some of them (in loops L2, L11, and L12) contribute to MT binding in a nucleotide state-dependent manner, giving rise to high and low affinity in the ADP and ATP states, respectively. Some of them (in loop L2 and K loop) contribute to MT binding in some kinesins (such as KIF1A) but not in others. Collectively, these charged residues may modulate the kinesin–MT binding affinity according to the functional roles of various kinesins. This study will offer promising targets for future mutational and functional studies for probing the motor mechanisms of kinesin and engineering of kinesins with novel motor properties.

ASSOCIATED CONTENT

Supporting Information

Breakdown of electrostatic ($\Delta\Delta G_{\text{elec}}$) and van der Waals ($\Delta\Delta G_{\text{VDW}}$) contributions to $\Delta\Delta G$ for selected charged residues in KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP} (Table S1) and electrostatic interactions involving the charged residues in key

secondary structural elements of KHC_{ADP}, KIF1A_{ADP}, and KIF1A_{ATP} (Figures S1–S3). This material is available free of charge via the Internet at <http://pubs.acs.org>.

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ABBREVIATIONS

MT, microtubule; MD, molecular dynamics; P_i, inorganic phosphate; EM, electron microscopy; KHC, human conventional kinesin; VDW, van der Waals; PME, particle mesh Ewald; PB, Poisson–Boltzmann; rms, root-mean-square; rmsd, root-mean-square deviation; MM-PBSA, molecular mechanics–Poisson–Boltzmann surface area; LIE, linear interaction energy method.

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