

Mutant alcohol dehydrogenase leads to improved ethanol tolerance in *Clostridium thermocellum*

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Clostridium thermocellum is a thermophilic, obligately anaerobic, Gram-positive bacterium that is a candidate microorganism for converting cellulosic biomass into ethanol through consolidated bioprocessing. Ethanol intolerance is an important metric in terms of process economics, and tolerance has often been described as a complex and likely multigenic trait for which complex gene interactions come into play. Here, we resequence the genome of an ethanol-tolerant mutant, show that the tolerant phenotype is primarily due to a mutated bifunctional acetaldehyde-CoA/alcohol dehydrogenase gene (*adhE*), hypothesize based on structural analysis that cofactor specificity may be affected, and confirm this hypothesis using enzyme assays. Biochemical assays confirm a complete loss of NADH-dependent activity with concomitant acquisition of NADPH-dependent activity, which likely affects electron flow in the mutant. The simplicity of the genetic basis for the ethanol-tolerant phenotype observed here informs rational engineering of mutant microbial strains for cellulosic ethanol production.

bioenergy | genomics | inhibitor | resequencing | 454

Fuels from cellulosic biomass are among the leading options to meet sustainability and energy security challenges associated with fossil fuels, and conversion processes featuring biological fermentation are among the leading options for producing cellulosic biofuels. Among fermentation-based conversion processes, use of cellulose-fermenting microorganisms without added enzymes—consolidated bioprocessing—has strong potential (1), and a variety of microorganisms are under development (2).

Clostridium thermocellum is a thermophilic bacterium that can rapidly solubilize biomass and use cellulose as a carbon and energy source. Wild-type (WT) strains produce ethanol as well as organic acids, but growth is inhibited by relatively low ethanol concentrations (<10 g/L; refs. 3 and 4). Cultures of *C. thermocellum* have been adapted to tolerate ethanol concentrations as high as 80 g/L (5), and although greater ethanol production has been reported for tolerant strains, the highest concentration of ethanol production reported for this organism is <30 g/L (6).

Ethanol tolerance and inhibition are typically complex, incompletely understood traits, although some studies point to compromised membrane integrity as a key factor (7, 8). In the case of *C. thermocellum*, Williams et al. (5) used MALDI-TOF analyses to obtain membrane proteomic profiles for WT and ethanol-adapted (EA) strains, which suggested that membrane-associated proteins were less abundant in EA strains or had issues related to incorporation into the cell membrane. Recently, Timmons et al. (9) showed that *C. thermocellum* EA had more fatty acids with chain lengths of >16:0 and significantly more 16:0 plasmalogens compared with WT and proposed that ethanol tolerance was due to fatty acid alterations that increased membrane rigidity to counteract the fluidizing effect of ethanol. However, the genetic basis for the enhanced ethanol tolerance for EA strains of *C. thermocellum* has not been determined. We hypothesized that the genome of *C. thermocellum* EA accumu-

lated one or more mutations that permitted its growth in higher concentrations of ethanol compared with the WT strain.

Results and Discussion

Resequencing Analyses. To determine the genetic basis of the *C. thermocellum* ethanol-tolerant phenotype, the genomes of EA and WT strains were resequenced by using a pyrosequencing approach (see Table 1 and *Materials and Methods* for details). A total of 72 and 500 454-pyrosequencing “high-confidence differences” (“HCDiffs”), as defined by the GSMapper software (454 Life Sciences), were detected for the WT and EA mutant (5), respectively (*SI Appendix, Tables S1 and S2*). The *C. thermocellum* WT and EA genomes were also resequenced by using a microarray-based comparative genome sequencing (CGS) approach, which revealed 410 putative differences (for details, see *Materials and Methods* and *SI Appendix, Table S3*). A summary of the combined pyrosequencing and CGS analyses is presented in Table 2 and *SI Appendix, Fig. S1*.

The number of putative differences identified by either pyrosequencing or CGS alone decreased dramatically as more stringent filtering criteria were applied (*SI Appendix, Fig. S2*). In contrast, most differences identified by both independent technologies were retained as confidence stringency was increased, and 230 genetic differences were validated by both resequencing platforms after filtering (*SI Appendix, Fig. S2B*). Repetitive DNA elements were identified in several lower-confidence differences and were filtered out (*SI Appendix, Fig. S3*), a finding that may point to areas for future bioinformatics improvements. Analysis of pyrosequencing data indicated that two large regions (~9 and 21.5 kb) encoding hypothetical or phage-related proteins were deleted in EA, and CGS data supported that these regions were deleted (*SI Appendix, Fig. S1 vii and viii*).

The ability to characterize strains through multiple high-throughput resequencing approaches not only provides rapid insights into adaptive evolution of strains with important biological or industrial traits, but also helps overcome the limitations associated with respective resequencing technologies. This study and previous pyrosequencing studies (10, 11) suggest that putative high-confidence differences with variation values of <80% should

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Table 1. Summary statistics for genome resequencing via GS FLX system

Genome	No. of reads	No. of bases	No. of contigs (>500 bp)	No. of assembled bases in contigs (>500 bp)	Genome coverage
<i>C. the</i> [†]	687,158	139,392,584	126	3,711,170	~38
<i>C. the</i> EtOH [‡]	451,118	105,817,828	122	3,677,593	~29

[†]*C. thermocellum* WT strain ATCC 27405.

[‡]*C. thermocellum* ethanol-tolerant mutant strain derived from ATCC 27405.

be interpreted with caution. We suggest that, although it is possible to rank putative differences based on possible technical limitations, lower-ranked differences cannot be disregarded entirely.

C. thermocellum EA SNPs were detected in genes that encode proteins previously reported as differentially expressed compared with the WT strain (*SI Appendix, Tables S2 and S3* and ref. 5). These genes included Cthe_0578 (glycoside hydrolase, family 9), Cthe_2263 (H⁺-transporting two-sector ATPase, C subunit), Cthe_1939 (magnesium transporter), Cthe_0858 (protein of unknown function DUF1432), Cthe_1385 (protein translocase subunit SecA), Cthe_1285 (metal-dependent phosphohydrolase), Cthe_3171 (S-layer domain-like protein), Cthe_2341 (glycosyl transferase, family 2), Cthe_0912 (glycoside hydrolase, family 10), Cthe_1284 (glycogen/starch synthases, ADP-glucose type), and Cthe_2664 (2-octaprenylphenol hydroxylase). In addition, mutations were detected at many other loci that did not have corresponding differences at the protein level and included important genes such as Cthe_0423 [encoding AdhE, bifunctional acetaldehyde-CoA/alcohol dehydrogenase (ADH)]. An analysis of distribution of the identified mutations across the *C. thermocellum* genome in terms of their type and the predicted operons and metabolic pathways was used to gain further insights into the 500 pyrosequencing differences.

Nonrandom Distribution of Mutations Across the Genome and Their Link to EA Phenotypes. The number of insertions, deletions, SNPs (synonymous and nonsynonymous), and multiple substitutions in coding and noncoding parts of the genome is presented in *SI Appendix, Fig. S4*. Single nucleotide substitutions are the dominant type of mutation that occurs in EA, and nonsynonymous substitutions were approximately twice as abundant as synonymous substitutions. Multiple substitutions, however, mainly targeted coding sequences of the genome, and insertions and deletions were overrepresented in noncoding sequences.

Further statistical analysis of the distribution of mutations revealed an increased number of mutations in EA at several locations, when considering the frequencies of different types of mutations, their total number, and the presence of neighboring genes or genes comprising an operon (*SI Appendix, Fig. S5*). A manual curation of the distribution pattern of the mutations identified 16 putative hot spots for mutation (Table 3). Genes linked to these hot spots included 72 out of 500 putative mutations identified (14%). Many EA membrane-associated proteins for carbohydrate transport and metabolism, including enzymatic and structural components of the organism's cellulosome, have been reported to be less abundant compared with the WT strain (5). It is therefore not surprising that 7 of the 16 mutation hot spots (hot spot IDs 1, 3, 6, 7, 9, 13, and 16) were in genes related to cellulose degradation, consistent with a previously observed poor growth phenotype for EA on crystalline cellulose (5). Poor growth of EA was confirmed, and acetate and ethanol were produced (*SI Appendix, Fig. S6*). Most of the hypothetical gene mutations found in hot spots in this analysis were adjacent to phage/transposase genes (Table 3).

Importantly, a putative operon (DOOR Database operon 295062; ref. 12) containing 10 adjacent genes (Cthe_0422 to Cthe_0431) likely involved in ethanol production was significantly

affected by different mutations (hot spot ID 9). The bifunctional acetaldehyde-CoA/ADH (AdhE; Cthe_0423) is an important enzyme for ethanol production and had two mutations, indicating an increased probability of importance for altered metabolism of EA. The EA AdhE contains two nonsynonymous mutations resulting in predicted amino acid changes. We pursued the hypothesis that this mutation may affect ethanol metabolism and therefore ethanol tolerance.

Mutant *adhE* Allele Alone Confers Increased *C. thermocellum* Ethanol Tolerance. To determine whether the mutant AdhE plays a role in EA's enhanced ethanol tolerance, the WT and mutant alleles of the *adhE* gene were cloned into a replicating plasmid and transformed into *C. thermocellum* DSM 1313 WT strain (i.e., *adhE*⁺). The resulting strains were then assayed for their ability to grow in medium with cellobiose as the sole carbon source and elevated levels of ethanol (Fig. 1). The strain carrying the mutant allele showed marked improvement in growth in the presence of 20 and 24 g/L added ethanol, and it was the only strain able to grow in the presence of 40 g/L added ethanol. Although 500 putative mutations were identified in EA (*SI Appendix, Table S2*), a single mutated gene from this culture was able to confer most of the ethanol-tolerant phenotype of EA (Fig. 1). However, the possibility that other mutations could also confer the EA phenotype cannot be ruled out. In this study, increased ethanol tolerance was not linked to higher ethanol yields (*SI Appendix, Fig. S6*), which is in keeping with a recent study that examined *Escherichia coli* isobutanol tolerance and productivity (13).

Plasmid DNA was unable to be isolated from cultures containing the mutant *adhE* allele grown with 40 g/L added ethanol during routine strain verification. However, when the chromosomal *adhE* was sequenced from this strain, a gene conversion event was discovered in which the mutant allele replaced the WT allele on the chromosome (*SI Appendix, Fig. S7*). This strain, which provides a clean genetic background for studying the effect of the mutant *adhE*, is herein called *C. thermocellum adhE*^{*}(EA).

Microbial ethanol tolerance has generally been thought to be a complex and likely multigenic trait (5, 7, 9). There have been suggestions that no single gene can endow microbes with tolerance to ethanol and other toxic compounds (14), and until recently little progress had been made in identification of key genetic changes that confer enhanced ethanol tolerance (15). Global transcription machinery engineering has been used to improve tolerance to both glucose and ethanol and to increase productivity in *Saccharomyces cerevisiae* by altering expression of

Table 2. CGS differences supported by pyrosequencing

CGS confidence values	No. of CGS differences	Subset supported by 454 [†]
High	234	230 (98%)
Medium	85	3 (4%)
Low	91	9 (10%)

[†]Variation values reported by the GSNMapper software are shown in parentheses. A higher score indicates more reads were in agreement, with the maximum value being 100.

Table 3. Mutational hot spots identified in the EA mutant genome

Hot spot ID	Genes with mutations	No. of mutations	Description [†]	Products
1	Cthe_2825 Cthe_2826	11	Two genes with multiple mutations in one TU	Hypothetical proteins
2	Cthe1806 Cthe1807	6	Two adjacent TU, two genes with INS	Cellulosome enzyme, dockerin type I
3	Cthe2018	6	One gene with multiple mutations	Hypothetical protein
4	Cthe0056 Cthe0059	5	Two adjacent TU and an intergenic regions	Ig-like, group 2 and cellulose-binding
5	the1235 Cthe1237	5	Two adjacent TU with INS	Cellulose 1,4-beta-cellobiosidase and leucyl-tRNA synthetase
6	Cthe1890	5	One gene with multiple mutations	Cellulosome enzyme, dockerin type I
7	Cthe2854 Cthe2855	5	Two genes with multiple mutations in one TU	Hypothetical proteins
8	Cthe3077 Cthe3078	4	Two genes with multiple mutations in one TU	Cellulosome anchoring protein, cohesin region
9	Cthe0422 Cthe0423	4	Two genes with two NON in each in one TU	Redox-sensing transcriptional repressor Rex and bifunctional acetaldehyde-CoA/ADH
10	Cthe2427	3	One gene with multiple mutations	Hypothetical protein
11	Cthe2559 Cthe2560	3	Two genes with STOP in one TU	DegT/DnrJ/EryC1/StrS aminotransferase and dTDP-4-dehydrorhamnose 3,5-epimerase
12	Cthe2611 Cthe2612	3	Two genes with two NON in one TU	Fibronectin, type III
13	Cthe1114 Cthe1115	3	Three genes in one TU with DEL	Tn7-like transposition protein C and HMG-I and HMG-Y, DNA-binding
14	Cthe0996 Cthe0997	3	Three total mutations and two NON including INS in two genes in one TU	DNA polymerase III PolC and 4-hydroxy-3-methylbut-2-en-1-yl diphosphate synthase
15	Cthe0659 Cthe0660 Cthe0661	3	Three adjacent genes and two NON including DEL	Hypothetical protein and glycoside hydrolase family protein
16	Cthe1885 Cthe1886	3	Two adjacent genes with INS	Phage integrase-like SAM-like and integrase catalytic subunit

[†]TU, transcription unit. Types of mutations are referred to as insertions (INS), deletions (DEL), and nonsynonymous substitutions (NON).

many genes simultaneously through a single genetic modification (16). In contrast, Hong et al. (15) used an inverse metabolic engineering approach to demonstrate that overexpression of en-

dogenous *S. cerevisiae* genes (INO1, DOG1, HAL1, or a truncated MSN2) individually can confer improved alcohol tolerance, higher titers, higher volumetric productivities, and increased

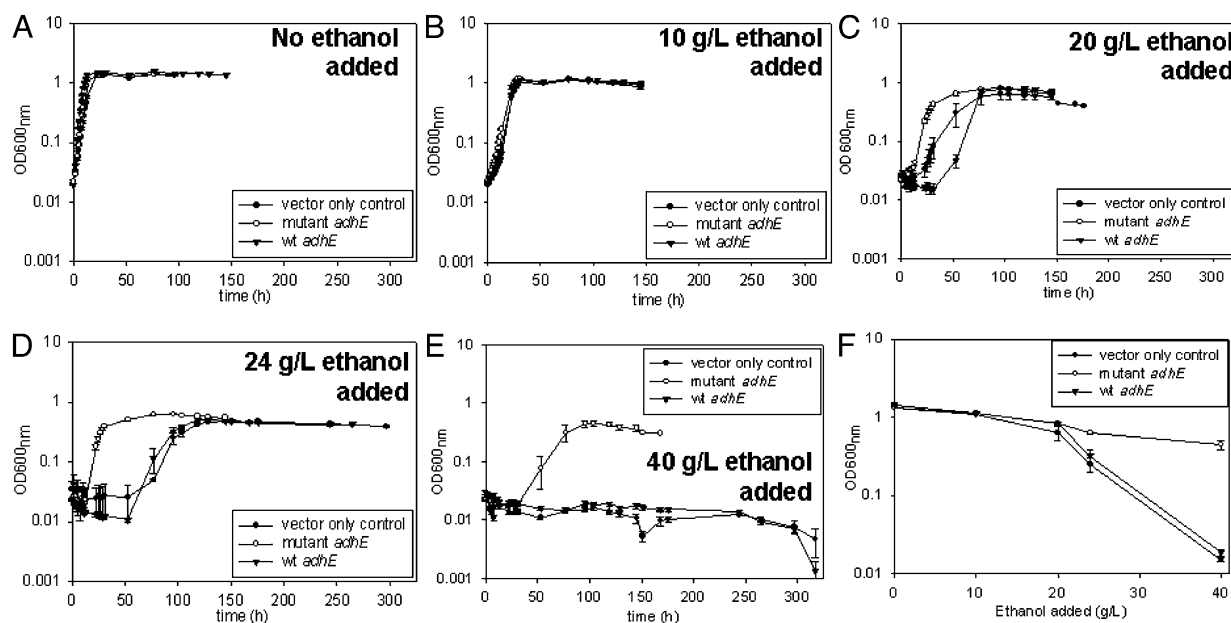


Fig. 1. Mutant *C. thermocellum* ADHs confers enhanced ethanol tolerance. (A–E) Growth of *C. thermocellum* DSM 1313 strains with different plasmids that provided a vector-only control, an additional copy of the WT version of *adhE*, and the mutant *adhE* gene was monitored by measuring culture turbidity (\log_{10} OD_{600nm}). Strains were grown at 55 °C in increasing amounts of added ethanol. (F) The final culture turbidities after 96 h of growth are presented as a function of ethanol added to the culture from the beginning the experiment. The graphs show the mean and the SE (bars) for three independent dose-response curves, and this experiment was repeated twice.

Although ethanol tolerance generally correlates with membrane alterations, ethanol tolerance may be limited by electron flux and central metabolism; however, further studies are required to examine carbon and electron flow, and possible linkages to changes in membrane composition. It is clear from this study that approaches to genetically modify *C. thermocellum* and possibly other microorganisms for biofuel production from cellulosic feedstocks must be reconsidered. Indeed, recent deletion of the *pta* gene, required for acetate production, resulted in the elimination of acetate as a fermentation end product but did not increase ethanol yield (26). Hence, not only ethanol tolerance but also ethanol production might be limited by electron flow as the ethanol concentration begin to rise. The use of a *C. thermocellum* strain with altered ADH cofactor specificity might help overcome issues related to carbon and electron flow. Aside from ethanol, the breadth of compounds tolerated by *C. thermocellum* strains EA and *adhE**(EA) is unclear. Future determination of compounds resisted by these strains may reveal the selective pressures that led to evolution of altered cofactor specificity of AdhE and suggest further paths for metabolic engineering of this organism for industrial biofuel production. Finally, the ability to identify and characterize sets of biological components linked to desired phenotypes, such as the mutated AdhE gene in this study, or overexpression of endogenous genes (11) offers the prospect for improved rational design of systems in the future that will be best suited to particular feedstocks and desired processes.

Materials and Methods

Detailed methods are available in the *SI Appendix, SI Materials and Methods*.

Strains and Culture Conditions. *C. thermocellum* WT strains ATCC 27405 and DSM 1313 were obtained from their respective culture collections. The *C. thermocellum* EA mutant culture was derived from strain ATCC 27405 and likely represents a mixed population of ethanol-tolerant strains. EA is the same culture that has been used in previous proteomics (5) and lipid (9) studies. The ethanol tolerance phenotype of EA has been described as being stable and retained after growth of >2,000 generations in the absence of ethanol.

Resequencing. Resequencing was conducted essentially as described (27). Briefly, genomic DNA from *C. thermocellum* WT ATCC 27405 or EA mutant cultures was sent to the NimbleGen facility for CGS service following the company's procedure. Pyrosequencing using the Roche 454 GS FLX System (454 Life Sciences) was carried out by using both shotgun and paired-end DNA library preparation methods, and sequences have been deposited in the National Center for Biotechnology Information (NCBI) Sequence Read Archive (accession nos. SRX030163.2 and SRX030164.1, respectively). The GSNapper application in the 454 GS FLX software package 1.1.03 (454 Life Sciences) was used to map the reads generated from GS FLX onto the *C. thermocellum* ATCC 27405 reference genome (GenBank accession no. CP000568).

Analysis of Distribution of Mutations Inferred from Pyrosequencing Data. Distribution of the mutations across the genome was analyzed by calculating quantities of each type of mutation (total mutations, nonsynonymous, synonymous, indels, and number of mutations in the intergenic regions) within each 1,000 and 5,000 nucleotides across the genome. For each position in the genome (L_i) we also calculated the local mutation frequency (LMP) as $LMP = 1/[(M_{up} - L_i) + (L_i - M_{down})]/2 * 100$, where L_{up} is the start position of the closest upstream mutation and L_{down} is the start position of the closest downstream mutation. Mutation hot spots were identified by manual curation of the distributions and by analysis of potential functional relationships between genes comprising each hot spot. In bacteria, the neighboring gene or genes may comprise an operon and are likely involved in the same biological process or metabolic pathway. To find out whether mutations target such functionally related genes, we calculated the number of mutations in each pair of genes and in genes that belong to an operon with subsequent manual analysis of the affected protein products and their annotation with MetaCyc pathways (28) using *C. thermocellum* pathway genome database available in the BioEnergy Science Center KnowledgeBase (http://cricket.ornl.gov/cgi-bin/beocyc_home.cgi). At least one nonsynonymous mutation in two adjacent genes or in a gene of the operon was required to consider mutations as functionally related. The intergenic region was considered as functionally related to a hot spot if the

distance from a mutation in the region to a mutated gene/operon was <400 nucleotides.

Plasmid and Strain Construction. The WT and EA *adhE* alleles were subcloned into plasmid pAMG205 (29), deleting *pyrF* and creating an artificial operon with the antibiotic resistance gene, *cat*, for expression in *C. thermocellum*. Empty vector control plasmid pAMG226 (pAMG205 Δ *pyrF*) was constructed via restriction digestion of pAMG205 (29) with *ZraI* and *SmaI*, followed by self-ligation of the 6.9-kb fragment and transformation into *E. coli* Top10. Plasmids were then transformed into *C. thermocellum* DSM 1313 through electroporation as described (29) with minor modifications.

ADH Enzyme Assays. The ADH enzyme assays were based on described methods (30, 31). Briefly, half-liter cultures of each strain were grown to $OD_{600} = 0.6$ and centrifuged at 4 °C at 6,000 \times g in a Beckman Coulter Avanti J-25 centrifuge with a JA-10 rotor. The culture was brought into the anaerobic chamber, and the supernatant was removed. All further steps were carried out in the anaerobic chamber. The cell pellet was resuspended in 4 mL of 100 mM Tris-HCl (pH 7.6)/0.1 mM DTT buffer, transferred to a 10-mL glass beaker, and sonicated for 8 min with 10-s pulses and 10-s pauses at 50% of the max intensity by using a Misonix Sonicator 4000 with a microtip. Crude cell extract was centrifuged at 14,000 \times g for 25 min and stored on ice until assayed. The anaerobic reaction mixture contained 100 mM Tris-HCl (pH 7.6)/0.1 mM DTT buffer, 0.5 mM NAD(P)H, 55 mM acetaldehyde, and 2–50 μ L of cell extract in 1.2-mL total volume. Decrease in absorbance was monitored at 340 nm to follow NAD(P)H oxidation (extinction coefficient 6.22 $\text{mM}^{-1}\cdot\text{cm}^{-1}$) by using an Agilent 8453 UV-Vis spectrophotometer with Peltier controlled heating set at 55 °C. Protein concentration was determined by using the Bradford method.

Homology Modeling of *C. thermocellum* AdhE. By using the protein sequence of the ADH (AdhE) domain of *C. thermocellum* ALDH/ADH, homology models of WT and double-mutant AdhE were constructed. The HHPRED webserver (32, 33), part of the Bioinformatics Toolkit webserver (34), was used to perform multiple sequence alignments of the *C. thermocellum* AdhE sequence to potential structural templates available in the PDB (35). The 2.7-Å resolution X-ray structure of 1,3-propanediol dehydrogenase from *Klebsiella pneumoniae* (PDB ID 3BFJ; ref. 36) and the 1.3-Å X-ray structure of Fe-containing ADH from *Thermotoga maritima* Tm0920 (PDB ID 1O2D; ref. 37) were selected as templates based on their homology to AdhE and inclusion of NAD(P), Fe, or both in the structures. The program MODELLER (38–41) was used to construct homology structures including the NAD cofactor and Fe ligand, and resulting models were assessed by using the Discrete Optimized Protein Energy (DOPE) method (42). The fully automated I-TASSER webserver (43) was also used to construct a homology model, and it was found that MODELLER produced structures with a more favorable (i.e., lower) DOPE score if the I-TASSER model was included as an additional template structure. To provide an additional level of validation of the homology structures, the MolProbity server (44, 45) was used to perform a Ramachandran analysis on the final models.

Molecular Dynamics (MD) Simulations. MD simulations were performed on the homology models on WT and double mutant (P704L, H734R) AdhE by using the program GROMACS (46) with the CHARMM 27 force field (47) and TIP3P water model (48). Previously reported Lennard-Jones parameters for Fe(II) were used (49). Energy minimization was performed on the homology structures by using the steepest descent method for 1,000 steps, and then each protein was solvated in a rectangular water box of with a minimum of 10 Å from the surface of the protein to the edge of the solvent box. Sodium cations were added to neutralize the charge of the system. Periodic boundary conditions were imposed, and the particle mesh Ewald method (50, 51) was used to describe long-range electrostatic interactions. MD simulations were carried out with an integration time step of 2 fs. To reach the target temperature (298 K) and pressure (1 bar), the Berendsen method was used with a relaxation time of 0.1 ps (52). After a 1-ns equilibration, production simulations were performed in the NPT ensemble by using the Nosé-Hoover thermostat (53, 54) and the Parrinello-Rahman barostat (55, 56) with relaxation times of 1.0 ps. The production run was carried out for 10 ns, and coordinates were saved every 1 ps for analysis.

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