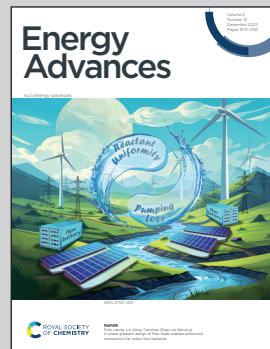


Showcasing research from Professor Shyue Ping Ong's laboratory, Department of NanoEngineering, University of California San Diego, USA.

Lithium dynamics at grain boundaries of β -Li₃PS₄ solid electrolyte

β -Li₃PS₄ (β -LPS) is a promising solid electrolyte for all-solid-state Li ion batteries due to its good Li ionic conductivity, good formability, and low processing cost. In this work, the Li ion transport in tilt and twist grain boundaries as well as amorphous/crystal interfaces of β -LPS was systematically investigated by performing large-scale molecular dynamics simulations with a highly accurate moment tensor interatomic potential. It was determined that the Li ion conductivities at the grain boundaries and amorphous/crystal interfaces are 1–2 orders of magnitude higher than that in the bulk crystal.

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Lithium dynamics at grain boundaries of β -Li₃PS₄ solid electrolyte[†]

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Lithium diffusivity at the grain boundaries of solid electrolytes (SEs) can strongly impact the final performance of all-solid-state Li ion batteries (SSLBs). β -Li₃PS₄ (β -LPS) is a promising SE due to its good formability and low processing cost, but its total Li ionic conductivity can vary by orders of magnitude depending on the synthesis and processing conditions. One of the possible sources for this variability, the GB contribution, is still poorly understood to date. In this study, we systematically investigate the Li ion transport in tilt and twist GBs as well as amorphous/crystal interfaces of β -LPS by performing large-scale molecular dynamics (MD) simulations with a highly accurate moment tensor interatomic potential (MTP). We find that the Li ion conductivities at the GBs and amorphous/crystal interfaces (10^{-4} – 10^{-3} S cm⁻¹) are 1–2 orders of magnitude higher than that in the bulk crystal (10^{-5} S cm⁻¹). The Li ion diffusivity at twist GBs and amorphous/crystal interfaces shows no correlation with the degree of diffusion isotropy, while tilt GBs exhibit a negative correlation. Using topological data analysis, the Li pathway network in twist GBs and amorphous/crystal interfaces comprises persisting large Li ring sub-networks ($n_{\text{Li}} \geq 5$) that closely resemble those found in the bulk amorphous structure, whereas more smaller and short-lived Li ring sub-networks ($n_{\text{Li}} \leq 5$) are detected in tilt GBs and the bulk crystal. The concentration of persisting large Li ring sub-networks in the GB and amorphous/crystal interfaces is directly proportional to the degree of Li site disordering which in turn correlates with GB conductivity. Our findings provide useful insights that can guide the optimization of conductivity not only in β -LPS but also in other sulfide-type solid electrolytes through possible GB engineering.

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Introduction

The strong interest in all-solid-state lithium-ion batteries (SSLBs) stems from their expected advantages, such as high safety, good durability, high energy density, and a simplified battery design for easy fabrication/manufacturing.¹ The high safety feature is mainly due to the use of an incombustible solid electrolyte (SE), a change from that of the combustible organic liquid electrolyte that is deployed in conventional Li ion

batteries.^{1,2} The use of a SE also enables achieving high energy density by using Li-metal or Li-alloy anodes and high capacity by using high-voltage cathodes.² Areas of application for SSLBs include electric vehicles and smart grids.^{1–3}

One active area of research in SSLBs is the development of high-performance and practical SE materials that demonstrate high room-temperature (RT) Li ionic conductivity ($\sigma_{300\text{K}} \geq 10^{-3}$ S cm⁻¹), crucial for high-rate battery charging/discharging.⁴ Amongst the various inorganic SE classes, sulfide SEs have high $\sigma_{300\text{K}}$ along with other desired SE characteristics such as easy processability and high mechanical malleability.^{5–8} The thiophosphate Li₃PS₄ (LPS) is a promising SE which exist in three known polymorphs: a low-temperature γ phase (*Pmn*2₁), a high-temperature β phase (*Pnma*) and a high-temperature α phase (*Cmcm*).^{8,9} In particular, β -LPS is known for its high $\sigma_{300\text{K}}$ (up to 10^{-4} S cm⁻¹) and facile synthesis.^{10–13} It has 2D zigzag Li diffusion pathways in the [010] crystallographic direction that consist of partially occupied 4b–4c Wyckoff site linkages (Fig. 1).¹³ Bulk Li ion diffusion is aided by cooperative PS₄^{3–} unit rotations.⁹

Apart from the bulk structural features, grain boundaries (GBs) and grain amorphization are other fundamentally

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† Electronic supplementary information (ESI) available: (i) GB structure generation, (ii) construction of GB seed structures of β -LPS for MTP training, (iii) construction of amorphous-crystal GB structures of β -LPS for use in MD runs with trained MTP, (iv) formation of topological ring structures by persistent homology, (v) MTP hyperparameter tuning details, (vi) MTP active learning results, (vii) MAE in energies and ion forces for bulk crystal and amorphous LPS by trained MTP, (viii) element-based MSD plots of GB models, (ix) MD-derived P and S heatmap at 300 K in GB and interface models, (x) diffusion behaviors at dislocation cores of GB structures, and (xi) visualization of Li ring sub-networks in GB structures. See DOI: <https://doi.org/10.1039/d3ya00234a>



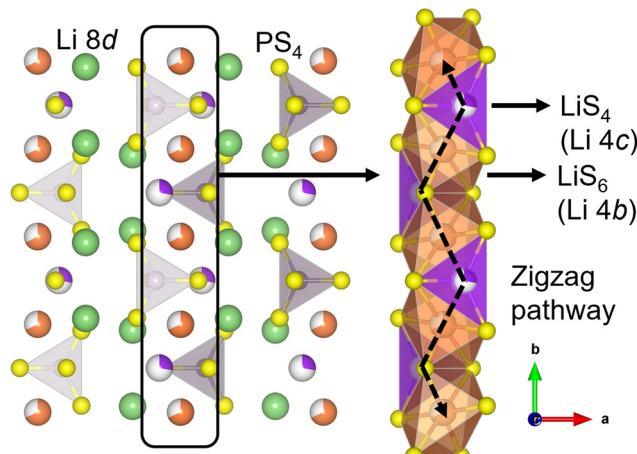


Fig. 1 Schematic picture for the *ab*-plane of β -Li₃PS₄ (spacegroup: *Pnma*). The isolated PS₄ unit, fully occupied tetrahedral Li 8d Wyckoff site, partially occupied octahedral Li 4b Wyckoff site, and partially occupied tetrahedral Li 4c Wyckoff site are shown in violet and yellow, green, orange with white and magenta with white, respectively. The zigzag 2D Li diffusion channels consist of alternating face-sharing LiS₆ and LiS₄ polyhedral units.

important microstructure-related features that affect overall σ_{300K} . The role that GBs play in affecting the electrochemical performance of SSLBs has been investigated in recent experimental works, particularly on issues related to leakage current, GB electronic conductivity, and Li dendrite formation.^{14,15} The measured σ_{300K} of β -LPS spans a large range between 10^{-7} and 10^{-4} S cm⁻¹. This may be directly related to GB resistance, aside from other factors related to synthesis conditions (e.g., phase purity, crystallinity, porosity, etc.).¹⁰⁻¹³ However, experimental studies of the relationship between GB structure and conductivity in β -LPS are challenging,^{16,17} especially since the Nyquist plot semi-arcs for bulk and GB contributions from impedance spectroscopy are often indistinguishable.^{10,12} Meanwhile, most computational studies so far have focused on bulk properties,¹⁸⁻²⁵ and only a very limited number of studies have investigated the effects of LPS microstructure towards Li ion transport. For example, MD simulations are performed by Ates *et al.* to study the particle size effect towards conductivity and by Kim *et al.* to probe the conductivity at the interface region formed between γ -LPS and its amorphous phase.^{26,27}

Machine learning interatomic potentials (MLIPs) that are trained using bulk and interface structures from AIMD simulations have been used to investigate various bulk and interface-related properties of SEs. For example, Wu *et al.* reported a combined experimental and computational study on Na_{3-x}Y_{1-x}Zr_xCl₆ SE, where MD simulations using a moment tensor potential (MTP) were employed to evaluate Na ionic conductivity, Na ion diffusion topology, and polyhedral rotation dynamics in the crystal structure.²⁸ Ko *et al.* also used MD with an MTP to investigate the improvement to the total Li ionic conductivity of a new class of high-entropy perovskite oxides through bulk compositional complexity and GB effects.²⁹ Finally, Wang *et al.* investigated by MTP-MD the atomistic

origin of Li ion transport resistance at the hetero-interface formed between SEs (LPS, Li₆PS₅Cl, and Na₃PS₄) and the pure (Li, Na) metal electrode.³⁰

In this work, the Li ion transport behavior at GBs and amorphous/crystal interfaces (to study amorphization effects) of β -LPS is systematically investigated by large-scale molecular dynamics (MD) simulations with a highly accurate MTP. Diffusivities, activation energies, and RT conductivities are estimated and compared for different GB (tilt and twist) and amorphous/crystal interface structures. The Li pathway network for GBs and amorphous/crystal interfaces is also analyzed by the persistent homology technique, specifically on the topological 1D Li ring features that are extracted from nanosecond-scale MD trajectories. Our findings provide new insights into the GB-region diffusion and conductivity not only in β -LPS but also in other sulfide-type SEs.

Methods

Moment tensor potential development

In recent years, MLIPs³¹⁻³⁵ have emerged as a means to accurately parameterize the potential energy surface (PES) of a collection of atoms using DFT calculations. For this work, we have selected the efficient MTP formalism, which has been extensively covered in prior works.³⁶⁻⁴⁰ Interested readers are referred to those for details. Here, we will focus on the workflow used for the development of the MTP, as shown in Fig. 2. For the investigations in this work, it is essential that the fitted MTP is able to capture the PES of both the bulk and GBs of the target β -LPS system accurately. The workflow is adapted from those developed previously by the current authors for the study of lithium superionic conductors and their interfaces,^{41,42} using the Python Materials Genomics (pymatgen) and Fireworks packages.^{43,44}

The MTP training is carried out in two stages. In the passive learning (PL) stage, two MTP hyperparameters – cutoff radius (r_c) and number of basis (n_{basis}) – are tuned using a grid search method with $r_c = [4.0 \text{ \AA}, 5.0 \text{ \AA}, 6.0 \text{ \AA}, 7.0 \text{ \AA}]$ and $n_{\text{basis}} = [8, 12, 14, 16, 18, 20]$. A total of 300 structures are randomly sampled from *ab initio* molecular dynamics (AIMD) simulations (300–2000 K) of the relaxed bulk β -LPS structure (Inorganic Crystal Structure Database ID: 35018).⁴⁵ The energy and force datasets are then each split in a 90:10 ratio into 270 training and 30 validation structures. Validation is performed based on mean absolute error (MAE) of total energies, forces, and stresses. Single-point density functional theory (DFT) calculations are then carried out on the structures to obtain accurate total energies and forces. All DFT calculations are performed using the Vienna *Ab initio* Simulation Package (VASP)^{46,47} within the projector augmented wave (PAW) method.⁴⁸⁻⁵⁰ The initial structure relaxations and AIMD simulations are carried out using the Perdew–Burke–Ernzerhof (PBE) generalized gradient approximation (GGA) functional,^{51,52} while the single point energy calculations are performed using the optB88 vdW functional.⁵³ As demonstrated in our previous work,⁴¹ such a



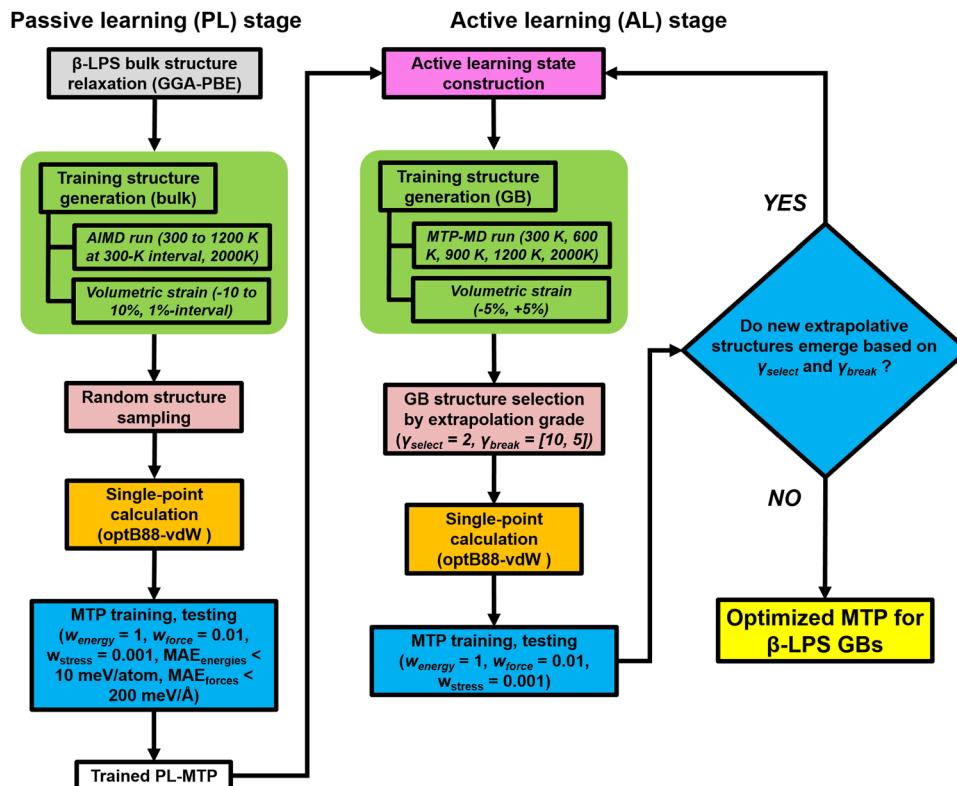


Fig. 2 Schematic workflow for passive and active learning related to MTP development for β -LPS GBs.

PBE-optB88 protocol captures the relevant energetics of sulfide SEs (especially the effect of non-local dispersion) at a minimal computational cost. The single point energy calculations are set with a kinetic energy cutoff and Gamma-centered k point grid of 520 eV and $1 \times 1 \times 1$, respectively. The convergence criteria for the electronic self-consistency loop, ionic relaxation loop, and forces are set to 1×10^{-5} eV, 2×10^{-2} eV, and 0.01 eV \AA^{-1} , respectively.

In the second active learning (AL) stage, the remainder of the MTP hyperparameters are optimized using additional isotropically-strained (-5% , $+5\%$) and GB structures. The GB structures are generated using the coincident site lattice (CSL) approach (see ESI[†]). We have adopted an extrapolation grade proposed previously to identify structures to be added to the training set.⁵⁴ The extrapolation grade threshold for structure selection (γ_{select}) is set to 2.0. Additionally, the threshold for termination (γ_{break}) is initially fixed to a looser setting ($\gamma_{\text{break}} = 10.0$) to promote structure diversity in the early AL iteration steps, and then gradually reduced ($\gamma_{\text{break}} = [5.0, 2.0]$) over the course of AL iteration steps. The purpose for this is two-fold: (i) to impose an increasingly stricter setting in the selection of extrapolative structures and (ii) to minimize the chance of unphysical structures from being picked during the later part of the AL stage. A separate test dataset outside of the AL loop is also generated, sampled from MD trajectories of 6 representative GB structures ($\Sigma 5[001]/(120)$ tilt, $\Sigma 5[010]/(010)$ twist, $\Sigma 3[110]/(111)$ tilt, $\Sigma 5[001]/(001)$ twist, $\Sigma 5[001]/(2-10)$ tilt, and $\Sigma 5[010]/(201)$ twist GB). The structures are chosen at regular intervals at different stages of MD calculations at a given temperature

(300, 800, 2000 K): (i) 5 structures from the heating stage, (ii) 5 structures from the equilibration stage, and (iii) 13 structures from the production stage. The DFT calculations and validation are carried out in the same way as the PL stage. The AL stage is terminated when no new structures emerge anymore based on the γ_{break} condition. Other GB structures are also considered during the actual deployment of the final MTP in MD production runs, these are $\Sigma 13[001]/(3-20)$ tilt, $\Sigma 13[001]/(150)$ tilt and $\Sigma 25[001]/(170)$ tilt GB. The atomic-scale configurations of different GB structures are provided in Fig. S1 (ESI[†]). The MLIP software package is used for MTP development.⁵⁴

Molecular dynamics simulations

The MD simulations are performed under the isothermal-isobaric (*NPT*) ensemble condition using the Large-scale Atomic/Molecular Massively Parallel Simulator (LAMMPS) package.^{55,56} Ion positions and variation in ion velocities are calculated by the velocity-Verlet algorithm.⁵⁷ The relaxation times of the Nosé–Hoover thermostat and the barostat are set to 0.1 and 1 ps, respectively.^{58,59} A 2 fs MD step size is used. The MD simulations on bulk, amorphous, GB, and amorphous/crystal interface structures are performed on supercells containing at least 10 000 atoms. Using the Einstein–Smoluchowski equation,⁶⁰ the diffusion coefficient (D) for Li ions is calculated from the slope of the mean square displacement (MSD) *versus* time (t) in the diffusive (linear) regime as follows:

$$D = \frac{\langle [\vec{r}(t + \tau) - \vec{r}(t)]^2 \rangle}{2dt}, \quad (1)$$

where $d = 3$ is the diffusion dimensionality, $\vec{r}(t)$ is the atom



position at time t , τ is the time lag between atom positions $\vec{r}(t + \tau)$ and \vec{r} , and $\langle [\vec{r}(t + \tau) - \vec{r}(t)]^2 \rangle$ is the MSD. The ionic conductivity is then given by the Nernst–Einstein equation:⁶¹

$$\sigma = \frac{ne^2 z^2 D}{k_B T}, \quad (2)$$

where n is the Li atom density, e is the elementary charge, z is the Li-ion charge, k_B is Boltzmann's constant, and T is the absolute temperature. Li ion diffusivity values at the GB region and the amorphous/crystal interface are extracted using a code that is developed by the current authors.^{62,63}

Amorphization of β -LPS

The β -LPS unit cell (12 Li, 4 P, and 16 S atoms; optB88-optimized lattice constants: $a = 13.07 \text{ \AA}$, $b = 8.13 \text{ \AA}$, $c = 6.26 \text{ \AA}$) is expanded into a $4 \times 8 \times 12$ supercell (12 288 atoms). This supercell is then used as input structure in the amorphization procedure by MTP-MD which comprises 5 stages: (i) heating up to 2000 K for 200 ps, (ii) equilibration at 2000 K for 400 ps, (iii) cooling down to 300 K with a rate of 5 mK fs^{-1} , (iv) equilibration at 300 K for 2 ns, and (v) production run at 300 K for 8 ns. The MD runs are performed under the *NPT* ensemble condition with a 2 fs MD step size. The relaxation times of the Nosé–Hoover thermostat and the barostat are set to 0.1 and 1 ps, respectively. The PRDF plots are derived from the MD trajectory of the production run stage.

Topological data analysis of the Li pathway network

The Li pathway topology at GB and amorphous/crystal interfaces is analyzed by persistent homology.⁶⁴ The procedure is schematically shown in Fig. S3 (ESI†). Starting from a set of Li ion coordinates, each point is replaced with a sphere with an initial radius (r) of zero. The radii of the spheres are then simultaneously increased. Ring “birth” occurs when the spheres touch each other forming closed ring structures. The sphere radius at which ring birth occurs is denoted as b . As the sphere radius is further increased, the ring eventually becomes completely filled, leading to ring “death”. The sphere radius at which ring death occurs is denoted as d . The record of sphere radii pairs related to each ring (b, d) is then compactly expressed into a topological object called the 2D persistence diagram (PD). A total of 200 MD snapshots of Li ion coordinates at the GB or amorphous/crystal interface regions are considered in this analysis. All the generated (b, d) pairs are then plotted, resulting to a 2D PD plot. PD analysis is performed using the scikit-TDA package.⁶⁵

Calculation of atom disorder parameter, visualization

The disorder parameter around each atom is calculated using the pascal package.⁶⁶ Visualization of crystal structures is made with VESTA software.⁶⁷ Plots are generated using the Matplotlib package.⁶⁸

Results

Accuracy of trained moment tensor potential

In the PL stage of MTP development, $r_c = 5.0 \text{ \AA}$ and $n_{\text{basis}} = 14$ are determined as the optimal hyperparameter combination

based on test-set MAE and MTP model complexity (Fig. S4 and S5, ESI†). These hyperparameter values are then employed in the subsequent AL stage. Fig. 3 shows the test-set MAE for a total of 1350 GB structures sampled from 6 test GB types and with MD runs at 300 K, 800 K, and 2000 K that are outside the AL loop. It can be observed that MAE in energies (e.g., Fig. 3a) converges to around 2.66 meV per atom (Fig. 3c) after 8 AL steps, while the MAE in forces (e.g., Fig. 3b) remains reasonably low from the start of the AL stage (i.e., $< 200 \text{ meV \AA}^{-1}$) and converges to a value of 147 meV \AA^{-1} for a total of 1 339 200 force components (Fig. 3d).

The final MTP is then further validated in terms of its ability to reproduce bulk and GB properties. As shown in Table 1, the MTP yields very low errors of 0.08–2.62% errors in lattice parameters, (ii) 0.43% error in crystal density, and (iii) the same order of magnitude for elastic moduli values. For the bulk amorphous phase, test errors also fall within the set cutoff values (Fig. S6, ESI†). The MAEs in the calculated GB energies are also extremely low at 0.10 J m^{-2} (Fig. 3e).

Partial radial distribution function analysis

Fig. 4 shows the partial radial distribution function (PRDF) plots of crystalline and amorphous β -LPS from MD simulations using the fitted MTP. For the crystalline β -LPS (Fig. 4a, left subplot),⁶⁹ Li ions are located in tetrahedral and octahedral sites. In the tetrahedral site, the Li–S distances are unequal due to the Li sitting closer to one of the tetrahedral faces. In the octahedral site, the 2 axial Li–S distances are longer than 4 basal Li–S distances. The Li–Li PRDF peaks can be assigned to the distances between Li in different Wyckoff sites ($8d$ – $8d$ $\sim 3.48 \text{ \AA}$, $8d$ – $4b$ $\sim 3.23 \text{ \AA}$ and $\sim 4.09 \text{ \AA}$, $8d$ – $4c$ $\sim 3.33 \text{ \AA}$, $4b$ – $4b$ $\sim 4.01 \text{ \AA}$). For the Li–P PRDF, the first peak at $\sim 3.05 \text{ \AA}$ mainly originates from Li $8d$ –P distance. For the Li–S PRDF, the first peak at $\sim 2.43 \text{ \AA}$ can be assigned to the tetrahedral LiS_4 coordination, the two shorter Li–S distances in octahedral LiS_6 , and the shortest Li–S distance in Li $4c$ – S_4 coordination. The first peak of the P–P PRDF at $\sim 5.01 \text{ \AA}$ is given by the separation between isolated PS_4^{3-} units. The average fraction of P^{5+} cation in a tetrahedral coordination with S^{2-} anions within a radius cutoff of 2.5 \AA is calculated to be 1 (or 100%). The S–S edge distance in the PS_4^{3-} unit is represented by the first S–S PRDF peak at $\sim 3.35 \text{ \AA}$.

The evidence for the structural change in the amorphized β -LPS can be seen in the decrease in the PRDF peak height and peak broadening for $r > 4 \text{ \AA}$ (Fig. 4a, right subplot). The presence of distinct PRDF peaks for $r < 4 \text{ \AA}$ indicates that short-range ordering is kept after the amorphization process. The appearance of new peaks for $r < 4 \text{ \AA}$ in the P–P PRDF can be assigned to P–S–P bridges, i.e., formation of condensed thiophosphate groups such as $\text{P}_2\text{S}_7^{4-}$, $\text{P}_2\text{S}_6^{2-}$, $\text{P}_3\text{S}_{10}^{5-}$, PS_4^{3-} , and P_2S_5 , as shown in Fig. 4b and c; the formation of these groups have concomitant S_2 species forming as well (see Fig. 4c). In contrast to crystalline β -LPS, only 87% of P^{5+} cations are in tetrahedral coordination while 13% form non- PS_4^{3-} units. The latter local structures are prevalent in amorphous Li_2S – P_2S_5 SEs.^{17,71}



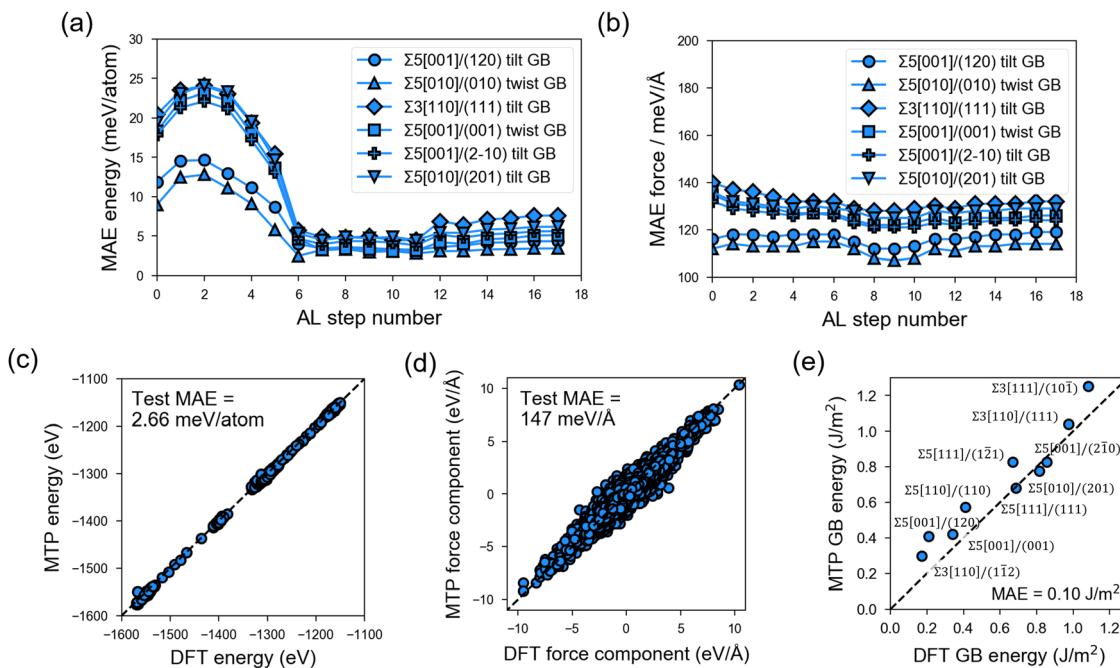


Fig. 3 Test-set accuracy of the trained moment tensor potential (MTP). Mean absolute errors (MAEs) in (a) energies and (b) forces for test β -LPS GB structures outside of the active-learning (AL) loop and sampled from 800 K on-the-fly MD runs (6 selected test GB types). Parity plots for MTP vs. DFT (optB88) (c) energies and (d) force components for test GB structures in (a) and (b) that are sampled from 300, 800, and 2000 K MTP-MD runs. (e) Parity plot for MTP vs. DFT GB energy. Dashed line indicates ideal $y = x$ relationship.

Li^+ diffusion barriers and conductivities

Fig. 5 shows the Arrhenius plots obtained from MD simulations of bulk, GB, and amorphous/crystal interfaces using the fitted MTP. Table 2 provides the extracted activation barriers (E_a) and σ_{300K} , including those from other computational and experimental studies for comparison. For GB/interface regions, a thickness of $\sim 10 \text{ \AA}$ for each of the 2 GBs/interfaces in the non-periodic direction of the structure models is considered for statistical sampling. The bulk β -LPS crystal (Fig. 5, top subplot) has a predicted $\sigma_{\text{bulk},300K}$ of $7.4 \times 10^{-5} \text{ S cm}^{-1}$ and an E_a of 0.29 eV. The predicted bulk crystal conductivity is about one order of magnitude smaller than that obtained from previous AIMD simulations using the PBE functional²⁴ and about 25% smaller than that experimentally measured *via* Li NMR.¹³ It should be noted that the PBE functional overestimates the lattice parameters of β -LPS, which would account for the overestimated diffusivities in the PBE-AIMD simulations. Furthermore, AIMD simulations are carried out at higher temperatures and for shorter time scales, which can result in further errors.⁴¹ The AC impedance measurements are for polycrystalline samples.

Table 1 Comparison between MTP, optB88 and experimental bulk crystal properties of β -LPS. Lattice parameters a , b and c ; density (ρ); bulk (B) and shear (G) moduli

Property	MTP	optB88 ⁴²	Expt. ^{69,70}
$a (\text{\AA})$	13.06	13.07	13.07
$b (\text{\AA})$	8.12	8.13	8.02
$c (\text{\AA})$	6.26	6.26	6.10
$\rho (\text{g cm}^{-3})$	1.864	1.860	1.872
$B (\text{GPa})$	30.5	22	10–12
$G (\text{GPa})$	9.8	11	5–6

There are two slope regimes in the present MTP-MD Arrhenius plot for the bulk crystal case, and this is consistent with observations based on temperature-dependent experimental conductivity measurements.^{10,12}

For the amorphous case (Fig. 5, top subplot), the predicted σ_{300K} of $1.1 \times 10^{-3} \text{ S cm}^{-1}$ is two orders of magnitude higher than in the crystalline bulk. The predicted σ_{300K} is close to the most recent measurement of σ_{300K} *via* AC impedance spectroscopy by Lu *et al.*⁷² though the experimentally reported σ_{300K} spans a wide range.^{17,73,74} Unlike crystalline β -LPS, no transition in Arrhenius regimes is observed, and the predicted E_a of 0.32 eV is also close to the experimental value by Lu *et al.*⁷²

Interestingly, tilt GBs are predicted to have σ_{300K} between that of crystalline and amorphous β -LPS, while twist GBs are predicted to have σ_{300K} comparable or slightly higher than that of amorphous β -LPS.

In the case of twist GBs and amorphous/crystal interfaces (Fig. 5, middle and bottom subplots), they have relatively lower $E_{a,GB}$ values which are in the range of 0.32–0.33 eV and comparable to the bulk amorphous structure ($E_{a,bulk} = 0.32 \text{ eV}$). Their σ_{300K} values are also in the same order of magnitude ($10^{-3} \text{ S cm}^{-1}$). For the amorphous/crystal interfaces, $E_{a,GB}$ does not significantly change with respect to the contacting crystal plane (e.g., (001), (100), and (120)), and the values are in good agreement with a previous classical MD result on amorphous/ γ -LPS interface (0.29 eV).²⁴

Li ion pathway and non-Li ion distribution at grain boundaries

The atomic distribution per element in the GB and amorphous/crystal interface structures are analyzed using their 300 K MTP-MD trajectories. Fig. 6a–c plot the heatmaps for Li atoms in

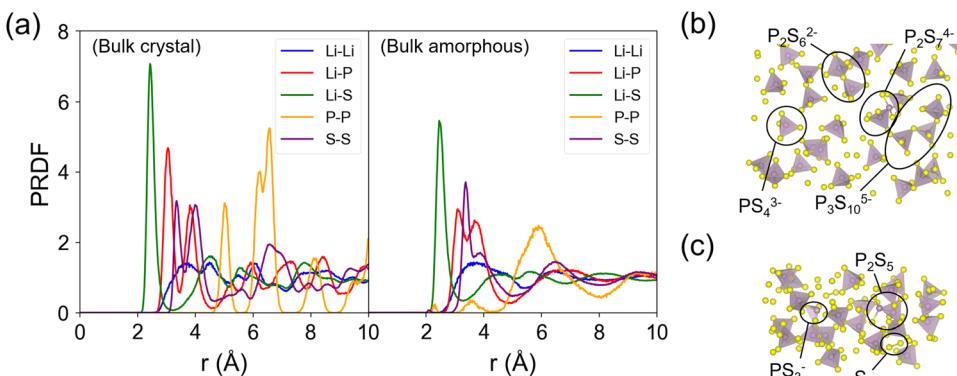


Fig. 4 (a) Partial radial distribution function (PRDF) plots of crystalline and amorphous β -LPS from MD simulations. The two PRDF subplots share the same vertical axis. (b) and (c) Representative local P–S structures in the amorphous β -LPS. Distinct host framework polyanion units are labeled; polyanion units are searched by incremental sweep of atom–atom distance and finding the minimum distance that results in polyhedral unit formation for all P atoms.

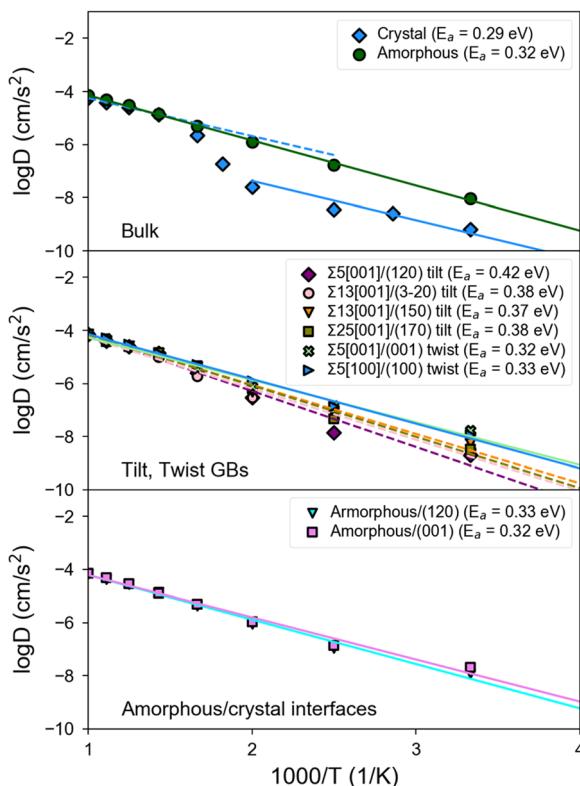


Fig. 5 Arrhenius plot of the log of the Li^+ diffusivity ($\log D$) versus inverse temperature ($1000/T$) in (top) bulk crystal and amorphous β -LPS; (middle) GBs and (bottom) amorphous/bulk interfaces of β -LPS. The $E_{a,\text{bulk}}$ value is fitted using the Arrhenius slope at the low temperature regime.

different tilt GBs. The diffuse Li atomic density at the GB region is clearly visible and extends up to several angstroms away from the GB center. The contiguous Li-density pattern hints at ion diffusion, which is also confirmed by the MSD plots (Fig. S7a–c, ESI†). For P and S atoms (Fig. S8, ESI†), their GB-region densities have point-like patterns just like the bulk region, indicating limited diffusion of these species. This was also

confirmed by their MSD plots (Fig. S7a–c, ESI†). In the case of twist GBs (Fig. 6d), a more extended and diffuse Li atomic density is observed (Fig. S7d, ESI†). In amorphous/crystal interfaces (Fig. 6e and f), highly diffuse Li density patterns are also observed.

Previous works have investigated the variation in Li ion concentration across GB regions in SEs, with the goal of trying to understand how GB structures affect the overall ionic conductivity.^{75,76} We performed a similar analysis and the results are shown in Fig. 7. The vertical axis represents the normalized Li count while the horizontal axis shows the distance in the GB- or interface-normal direction. In Fig. 7(a)–(c), tilt GBs have similarly shaped troughs and valleys as with the bulk region. In comparison, twist GBs (Fig. 7d) have Li ion concentration profiles that are flattened out, suggesting a significant change in the spatial distribution of Li sites and the number of accessible Li sites relative to the bulk region. This points to a change of the GB Li pathway network, which is likely caused by GB-region structure disordering and atomic rearrangement to lower the GB energy. A similar behavior is also noted in amorphous/crystal interfaces (see Fig. 7e and f). Meanwhile, investigated GBs and amorphous/crystal interfaces do not show Li depletion, which means that there is a low likelihood of GB- and interface-region Li trapping that can reduce Li ion diffusion. The GB- and interface-region disordering will be further analyzed in the later section.

We analysed the Li diffusion behavior near the misfit dislocation core of GB structures using trajectories from 300 K MD simulations. Two tilt GB structures ($\Sigma 55[001]/(120)$ tilt GB ($\alpha_1 = 53.13^\circ$), $\Sigma 5[001]/(170)$ tilt GB ($\alpha_1 = 16.26^\circ$) and one twist GB structure ($\Sigma 55[001]/(001)$ twist GB ($\alpha_2 = 180^\circ$))) were investigated. A 3D-box boundary region within the GB regions is specified as the dislocation core region (Fig. S9a–c, ESI†). It is found that the MSD of Li within the dislocation cores is generally comparable to that in the respective GB regions (Fig. S9d–f, ESI†). Hence, the dislocation core can be considered as a spatial region with an active Li ion pathway and an important feature that describes the overall GB-region Li diffusivity of GB structures.

Table 2 Calculated and experimentally measured E_a and σ_{300K} of bulk, GB, and amorphous/crystal interface structures of β -LPS at room temperature (300 K)

Structure	Ref.	Method	E_a (eV)	σ_{300K} (S cm $^{-1}$)
Crystalline β -LPS	This work	MTP-MD	0.29	7.4×10^{-5}
19	Morse-MD	0.41	—	
24	PBE-AIMD	0.35	8.8×10^{-4}	
13	Li NMR	0.24	1.0×10^{-4}	
27	AC impedance	0.29	4.9×10^{-5}	
10	AC impedance	0.47	3.0×10^{-7}	
Amorphous LPS	This work	MTP-MD	0.32	1.1×10^{-3}
26	Morse-MD	0.28	4.1×10^{-4}	
20	PBEsol-AIMD	—	8.8×10^{-5}	
72	AC impedance	0.26	1.3×10^{-3}	
17	AC impedance	0.40	2.8×10^{-4}	
73	AC impedance	0.35	2.0×10^{-4}	
74	AC impedance	0.41	1.2×10^{-4}	
Nanoporous β -LPS	12	Li NMR	0.36	1.6×10^{-4}
$\Sigma 5[001]/(120)$ tilt GB	This work	MTP-MD	0.41	3.1×10^{-4}
$\Sigma 5[001]/(130)$ tilt GB	This work	MTP-MD	0.42	1.1×10^{-4}
$\Sigma 13[001]/(150)$ tilt GB	This work	MTP-MD	0.37	6.5×10^{-4}
$\Sigma 13[001]/(3-20)$ tilt GB	This work	MTP-MD	0.38	3.6×10^{-4}
$\Sigma 25[001]/(170)$ tilt GB	This work	MTP-MD	0.38	3.9×10^{-4}
$\Sigma 29[001]/(-250)$ tilt GB	This work	MTP-MD	0.36	7.6×10^{-4}
$\Sigma 5[001]/(001)$ twist GB	This work	MTP-MD	0.32	2.0×10^{-3}
$\Sigma 5[100]/(100)$ twist GB	This work	MTP-MD	0.33	1.4×10^{-3}
Amorphous/(120)	This work	MTP-MD	0.33	1.4×10^{-3}
Amorphous/(001)	This work	MTP-MD	0.32	2.4×10^{-3}

Direction-dependent diffusivity at grain boundaries

Fig. 8 displays the direction-dependent Li-ion MSD plots (x and y for GB- or interface-lateral and z for GB- or interface-normal direction) for selected GBs from MTP-MD runs at 300 K. In $\Sigma 5[001]/(120)$ ($\alpha_1 = 53.13^\circ$) and $\Sigma 5[001]/(170)$ ($\alpha_1 = 16.26^\circ$) tilt GBs (Fig. 8a, left and right subplot, respectively), the MSD slopes along the x , y , and z directions are found to be

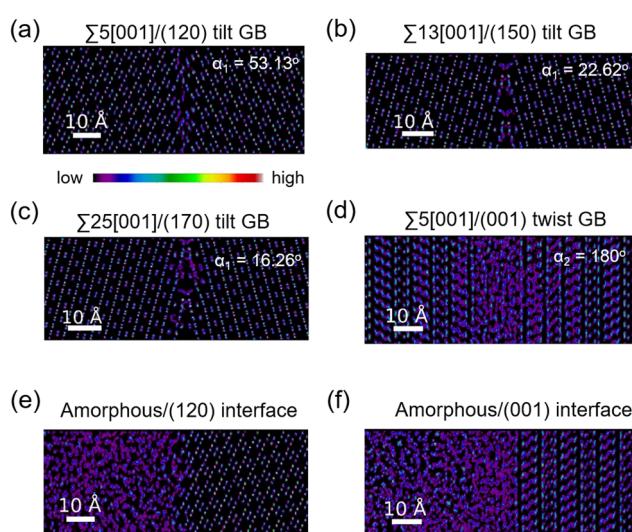


Fig. 6 2D-projected atom-count heatmaps for Li atoms from different GB structures as derived from NPT-MD calculations at 300 K: (a) $\Sigma 5[001]/(120)$ tilt GB, (b) $\Sigma 13[001]/(150)$ tilt GB, (c) $\Sigma 25[001]/(170)$ tilt GB, (d) $\Sigma 5[001]/(001)$ twist GB, (e) amorphous/(120) interface, and (f) amorphous/(001) interface. The length scale in the images is shown by the horizontal white line (10 Å). Tilt and twist angles are indicated as α_1 and α_2 , respectively.

comparable. Meanwhile, in $\Sigma 5[001]/(150)$ ($\alpha_1 = 22.62^\circ$), the MSD slope in the z -direction is shown to be smaller than in the x - and y -directions, suggesting that GB-normal diffusion is relatively slower than GB-lateral diffusion (*i.e.*, anisotropic diffusivity). Anisotropy is also noted in $\Sigma 5[001]/(001)$ twist GB ($\alpha_2 = 180^\circ$) but not in the amorphous/(120) interface structure, see Fig. 8b. The direction-dependent GB-region diffusivity has been suggested as well in Li₃N SE GBs.⁷⁷ Here, a simple parameter for the degree of diffusion isotropy (ξ) is derived using the ratio of the MSD slope components:

$$\xi = m_z / \max(m_x, m_y), \quad (3)$$

where m represents the slope of the MSD component in one GB or interface direction. This definition implies that when ξ is close to 1, the GB or interface region is highly diffusion-isotropic, otherwise diffusion anisotropy becomes dominant. Fig. 8c shows the total- $\log(D_{300K})$ vs. ξ plot for the different investigated GBs and amorphous/crystal interfaces. As can be seen, there is a negative correlation between the two quantities for tilt GBs, while for twist GBs and amorphous/crystal interfaces there is no correlation observed. Twist GBs and amorphous/crystal interfaces have consistently higher diffusivity than tilt GBs.

The observed trends for direction-dependent diffusivities of the different GB structures can be analyzed together with the 2D-projected MTP-MD atom trajectory densities. In the case of $\Sigma 5[001]/(120)$ tilt GB, which has a relatively large ξ but low D amongst the investigated tilt GBs, it exhibits a limited extent of spatial P disordering which means that the framework structure has less disordering as well (*i.e.*, only spanning for 2 P atomic layers in the GB-normal direction, see Fig. S8, ESI[†]). This narrow region of P disordering has a highly localized Li

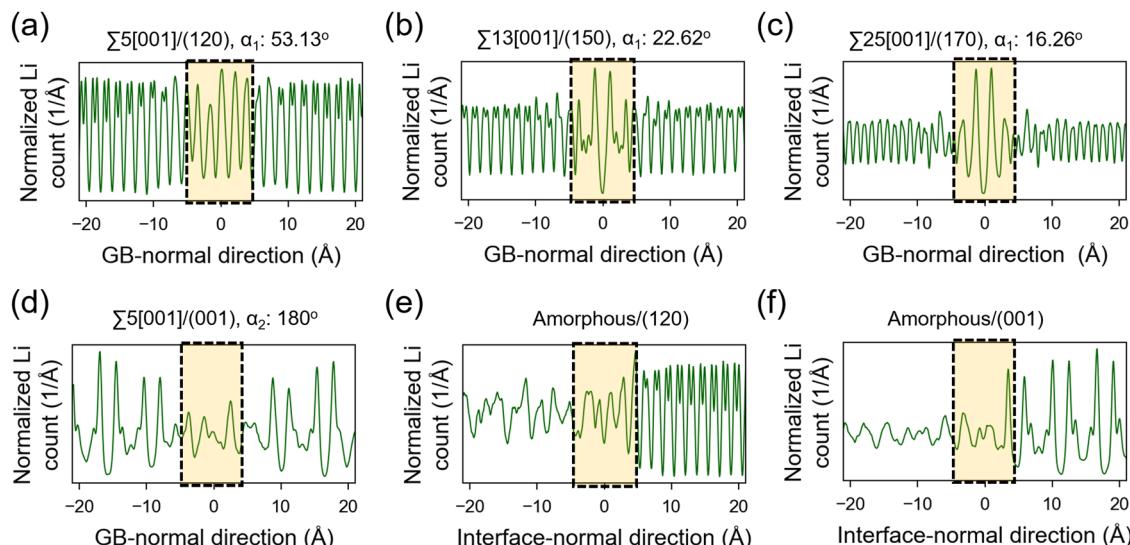


Fig. 7 Projected Li concentration profiles around the grain boundary (GB) and amorphous/crystal interface regions: (a) $\Sigma 5[001]/(120)$ tilt GB, (b) $\Sigma 13[001]/(150)$ tilt GB, (c) $\Sigma 25[001]/(170)$ tilt GB, (d) $\Sigma 5[001]/(001)$ twist GB, (e) amorphous/(120) interface, and (f) amorphous/(001) interface. Enclosed regions highlight the GB region. A Gaussian kernel smoothing procedure is performed with a covariance factor of 1.0×10^{-4} .

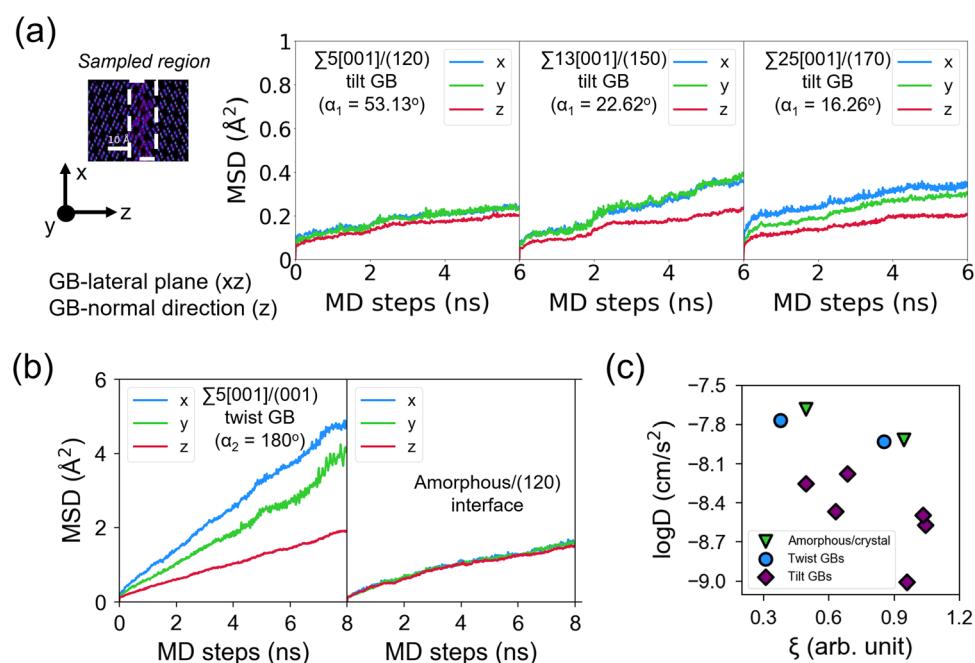


Fig. 8 Direction-dependent Li-ion mean squared displacement (MSD) component plots for selected β -PS GB and amorphous/crystal interface structures from 300 K MD runs: (a) tilt GBs: $\Sigma 5[001]/(120)$ tilt GB (left subplot), $\Sigma 13[001]/(150)$ tilt GB (middle subplot), and $\Sigma 25[001]/(170)$ tilt GB (right subplot); (b) twist GBs: $\Sigma 5[001]/(001)$ twist GB (left subplot), amorphous/(120) interface (right subplot). MSD components are determined from 300 K MTP-MD trajectories, with x and y representing GB- or interface-lateral directions and z representing GB- or interface-normal direction. Inset in (a) shows the corresponding 2D-projected Li-ion heatmap, with highlighted GB regions (enclosed in dashed lines) from where the sampling for MSD components is performed. A GB or interface region thickness of $\sim 10 \text{ \AA}$ with respect to the z -direction is used. (c) Plot of total- $\log(D_{300K})$ as a function of Li diffusion isotropy parameter ξ (eqn (3)).

atomic density and oriented pathway connectivity across the GB interface (see Fig. 6a), indicating that the GB Li pathway has largely retained the character of the β -LPS bulk Li pathway. Previous calculation work has shown that there is a specific

crystal direction in the β -LPS bulk (*i.e.*, [010]) that has a low Li ion migration barrier, while the other orthogonal directions ([100] and [001]) are predicted to have relatively higher barriers.⁷⁸ Given this, the favorable Li pathway relative to the



bulk may be blocked depending on the GB crystal orientation. In $\Sigma 5[001]/(120)$ tilt GB, the GB crystal orientation (*i.e.*, (120)) and the less disordered framework suggest that the relatively low D is mainly due to the significant contribution of the diffusivity component characterized by the less energetically favorable bulk-like Li pathway (*i.e.*, [100] direction of bulk). On the other hand, the $\Sigma 5[001]/(001)$ twist GB, which has a relatively low ξ value but high D , has a relatively large P disordering region (*i.e.*, 5–6 P atomic layers) around the GB interface (see Fig. S8, ESI[†]). This larger GB-region P disordering and the diffused GB-region Li atomic density (see Fig. 6d) are evidence of new and energetically accessible local pathways for Li ions, thus explaining the high D value. The favorable direction for Li ion migration relative to the bulk (*i.e.*, [010] direction) is found along the GB-lateral direction. A similar observation is found in amorphous-crystal interfaces which are determined to have both high ξ and high D . The large GB-region P disordering from the amorphous slab side induces the formation of more accessible local Li pathways across the GB interface and towards the crystalline side (see Fig. 6e and f). Overall, the disordering of the crystal structure framework is one of the key factors for unlocking high diffusivity and high pathway connectivity in β -LPS GBs. This is rather intuitive because disordering is often associated with defect formation (*e.g.*, Li vacancies, Li interstitials, and Li/P/S Frenkel-like defects) and chemical bond reorganization, both can change the inherent bulk anisotropic diffusivity behavior of β -LPS to a more isotropic one by increasing the number of accessible local diffusion routes at the GB region. Moreover, the present MTP-

MD results suggest that the modulation of GB crystal orientation in (highly) crystalline tilt GBs can only change the magnitude of D by one order at most. To provide more insight and description into the diffusivity behavior and local structure of different β -LPS GBs, further analyses are performed in succeeding sections.

Topology of the Li pathway at the grain boundaries

The GB Li pathway network is studied using the Li-ring persistence diagram (PD) which is constructed by applying the persistence homology technique⁶⁴ on the GB-region 300 K MD trajectories of Li ions. The Li ring sub-networks at the GB region can be visualized as edge connections forming as the Li–Li cutoff distance ($d_{c,Li-Li}$) is increased. Examples are shown in Fig. S10 and S11 (ESI[†]) for $\Sigma 5[001]/(120)$ tilt GB and $\Sigma 5[001]/(001)$ twist GB, respectively. In the PD approach, the one-half value of $d_{c,Li-Li}$ can be related to the filtration level for ring birth (b) when the Li sphere radius is increased until neighboring Li spheres touch each other and form a closed loop or a ring feature (*i.e.*, $b = 0.5d_{c,Li-Li}$). When the radius of the Li spheres is further increased until the ring hole is filled, ring death (d) occurs. The lifetime (L) of a Li ring feature can be calculated as $L = d - b$. Physically, L measures the closeness of neighboring Li ions in a ring. As a simplified case, a ring that is formed by Li ions in proximity will have a small L (*i.e.*, short-lived feature), while a ring that is formed by Li ions in proximity except for one that is located relatively far from the rest will have a large L (*i.e.*, long-lived or persisting feature). Therefore, L positively correlates with the distance of the farthest Li ion(s) in the ring.

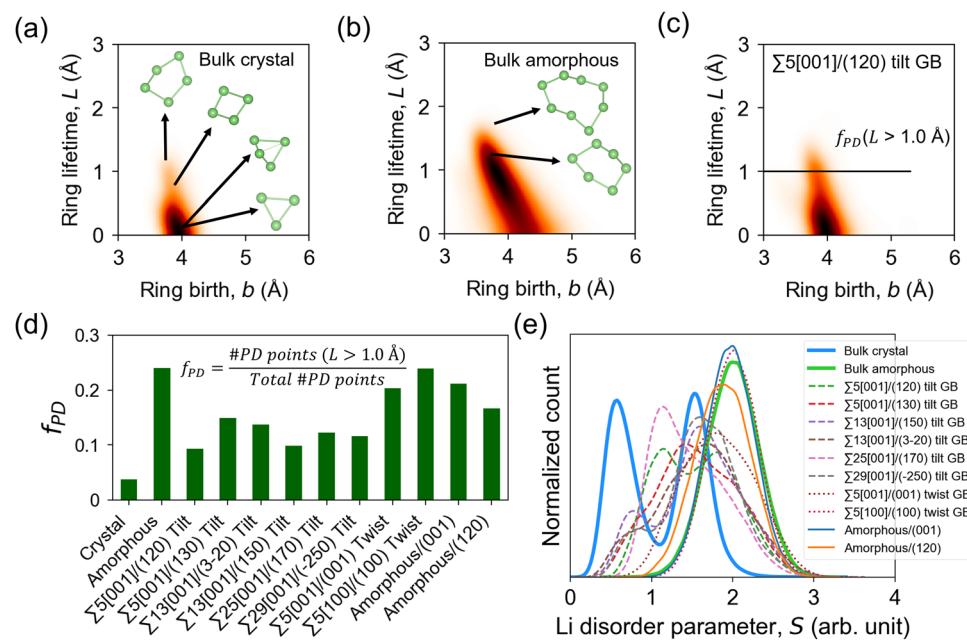


Fig. 9 Persistence diagrams (PD) of 1D topological ring features from MD-snapshot Li ion coordinates at 300 K: (a) bulk β -LPS crystal, (b) bulk amorphous phase, (c) $\Sigma 5[001]/(120)$ tilt GB. Heatmaps are smoothed by a Gaussian filter with a parameter $\sigma_{\text{smooth}} = 6$ Å; a higher PD density region is shown in a relatively darker color. Representative Li ring features of different sizes from the PD plots are shown in (a) and (b). (d) Calculated fraction of PD ring features (f_{PD}) with lifetime $L > 1.0$ Å ($L = d - b$, where d and b are ring death and ring birth times, respectively) and (e) distribution of Li site disorder parameter values (eqn (4)) for different bulk, GB, and amorphous/crystal interface structures.



PD analysis has been successfully used in the local structure analysis of amorphous solids, which are intrinsically disordered,^{79,80} similar to GB structures.

Fig. 9a–c shows the PD heatmaps of bulk, GB, and amorphous/crystal interface structures. For the bulk crystal (Fig. 9a), the highest concentration (darkest area) is located near the *b* axis at *b* ~ 4 Å. These features are generated from short-lived Li rings, such as by 3-Li rings, which are formed from two Li 4b and one Li 4c sites (Fig. 1). Larger and more persistent rings, such as 4-Li and 5-Li rings, are found at *b* < 4 Å and *L* > 1.0 Å. In the case of the bulk amorphous phase (Fig. 9b), the high-density region is shifted away from the *b* axis. Even larger rings are also detected, which include the more persistent 6-Li and 8-Li rings in the region defined by *L* > 1.0 Å and the short-lived ones in the region bounded by *b* > 4.5 Å and 0 < *L* < 1.0 Å. To quantitatively evaluate the topological difference in the Li pathway network in bulk, GB, and amorphous/crystal interface structures, we calculate the fraction of PD points (f_{PD}) in the PD region with *L* > 1.0 Å. The f_{PD} value of the bulk amorphous structure is calculated to be 4.6× larger than the bulk crystal. This comparison hints at the topological difference in the Li pathway network between the two bulk phases and may also partly explain the difference in their calculated *D*, *E_a*, and σ_{300K} values. For tilt GBs, a qualitative similarity of the PD shape as with the bulk crystal phase is noticeable (e.g., $\Sigma_5[001]/(020)$ in Fig. 9c), especially on the location of the PD high-density center near the *b* axis. Their main difference is the 2.7× larger f_{PD} value of the former than the latter. The f_{PD} values for the rest of the GB and amorphous/crystal interface structures are plotted in Fig. 9d, the order for the average f_{PD} (\bar{f}_{PD}) values is given as follows: $\bar{f}_{PD,crystal}(0.04) < \bar{f}_{PD,tilt}(0.12) < \bar{f}_{PD,twist}(0.22) \approx \bar{f}_{PD,amorphous/crystal}(0.19) \approx \bar{f}_{PD,amorphous}(0.24)$. This suggests that the f_{PD} parameter is related to the degree of structure disordering.

We compute the average disorder parameter around each Li ion (S_j , where *j* is the central Li ion) as follows:⁸¹

$$S_j = (1/n_{\text{neighbors}}^j) \sum_{k \in \text{neighbors}} [S_{jj} + S_{kk} - 2S_{jk}], \quad (4)$$

where S_{jj} and S_{kk} are complex numbers (q_{lm}^j, q_{lm}^k) described by the product of the spherical harmonics function $Y_{lm}(r)$ of degree *l* and vector r_{jk} , while S_{jk} is the scalar product given by $S_{jk} = \sum_{(-l \leq m \leq l)} q_{lm}^j (q_{lm}^k)^*$; *l* is set to 6. Fig. 9e plots the S_j distributions for bulk, GB, and amorphous/crystal interface structures. A bimodal distribution for the bulk crystal is revealed which can be assigned to the tetrahedral Li 8d/4c sites and the octahedral Li 4b sites. The distribution becomes unimodal for GB and amorphous/crystal interface structures, with the peak shifting towards larger S_j . The same general order is found for average S_j (\bar{S}_j) as with \bar{f}_{PD} : $\bar{S}_{j,crystal}(1.25) < \bar{S}_{j,tilt}(1.67) < \bar{S}_{j,twist}(1.94) \approx \bar{S}_{j,amorphous/crystal}(2.01) \approx \bar{S}_{j,amorphous}(2.00)$. The order for \bar{S}_j (and \bar{f}_{PD}) also corresponds well with MD-predicted σ_{300K} : bulk crystal ($10^{-5} \text{ S cm}^{-1}$) < tilt GBs ($10^{-4} \text{ S cm}^{-1}$) < twist GBs ($10^{-3} \text{ S cm}^{-1}$) \approx amorphous/crystal interfaces ($10^{-3} \text{ S cm}^{-1}$) \approx bulk amorphous phase ($10^{-3} \text{ S cm}^{-1}$). The

enhanced local disorder has been proposed to be beneficial as well in the $\text{Li}_{10}\text{GeP}_2\text{S}_{12}$ SE.⁸²

Conclusion

To conclude, we have performed a comprehensive study of the different GBs and amorphous/crystal interfaces in β -LPS using MD simulations with a newly developed MTP. An active learning protocol systematically samples the poorly represented regions of the potential energy surface, resulting in an MTP that can reliably reproduce key structural features and properties of bulk and amorphized β -LPS. We find that the lithium ionic conductivities in the GBs and amorphous/crystal interface in β -LPS are generally higher than the bulk crystal by 1–2 orders of magnitude. Thus, we surmise that the lower total conductivity observed in some previous experimental works may thus be related to other factors, such as final stoichiometry/composition, stabilized crystalline phase, and porosity. The predicted high σ_{300K} of $2.0 \times 10^{-3} \text{ S cm}^{-1}$ for the $\Sigma_5[001]/(001)$ twist GB structure, which is also determined to have a highly isotropic diffusivity, provides a potential target for GB engineering in β -LPS. We also find that the lithium ionic conductivity is correlated with structure disorder. Experimentally, disordering is often introduced by amorphizing the SE crystalline phase and it is closely linked to achieving an open structure and a large free volume that facilitates fast ionic conduction (e.g., in glass-ceramic SEs).⁸³ The tilt GBs, twist GBs, and amorphous/crystal interfaces in β -LPS possess varying degrees of amorphous-like character which, in this work, has been quantified in terms of the concentration of large and persisting local Li ring features in the Li GB pathway network and the GB-region Li site disordering. Therefore, control of such GB features can be a key towards an effective optimization of SE conductivity in β -LPS.

Conflicts of interest

There are no conflicts to declare.

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