## Structure and Dynamics of the Molten Alkali-Chloride Salts from an X-ray, Simulation, and Rate Theory Perspective

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Structure and Dynamics of the Molten Alkali-Chloride Salts from an X-ray, Simulation, and Rate Theory Perspective†

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Molten salts are of great interest as alternative solvents, electrolytes, and heat transfer fluids in many emerging technologies. The macroscopic properties of molten salts are ultimately controlled by their structure and ion dynamics at the microscopic level and it is therefore vital to develop an understanding of these at the atomistic scale. Herein, we present high-energy X-ray scattering experiments combined with classical and ab initio molecular dynamics simulations to elucidate structural and dynamical correlations across the family of alkali-chlorides. Computed structure functions and transport properties are in reasonably good agreement with experiments providing confidence in our analysis of microscopic properties based on simulations. For these systems, we also survey different rate theory models of anion exchange dynamics in order to gain a more sophisticated understanding of the short-time correlations that are likely to influence transport properties such as conductivity. The anion exchange process occurs on the picoseconds time scale at 1100 K and the rate increases in the order KCl < NaCl < LiCl, which is in stark contrast to the ion pair dissociation trend in aqueous solutions. Consistent with the trend we observe for conductivity, the cationic size/mass, as well as other factors specific to each type of rate theory, appear to play important roles in the anion exchange rate trend.

1 Introduction

Molten salts are resurfacing as integral components of clean, sustainable, and energy-efficient technologies. Their utilization as electrolytes in advanced intermediate- and high-temperature (>200 °C) batteries facilitates attaining high energy density and high power density typically required for electric vehicles and concentrating solar power plants. Molten salts can operate as coolants in nuclear reactors at high temperatures (500 °C-900 °C) but near ambient pressure, ensuring efficient and safe operation of the reactors. Furthermore, different eutectic mixtures of molten salts are promising solvents for electrorefining of nuclear wastes, chemical separation of actinides and rare earths, metal production, alloy heat treatment, and electrochemical synthesis of carbonaceous materials. Enabling this plethora of potential applications requires in-depth understanding of the correlation between microscopic structure, dynamics and macroscopic transport properties such as diffusivity and conductivity.

This article focuses on the alkali-chloride family of molten salts for which we make predictions on what the important driving forces are for transport. The accuracy of these predictions strongly depends on the quality of our simulations. For example, the 2D free energies that will be introduced in our rate theory study are directly related to the structural distributions of ions in different coordination shells. A comparison between computed and measured X-ray/neutron scattering data including structure functions and concomitant pair distribution functions (PDFs) can help validate the accuracy of models and simulations in determining molten salt coordination environments and associated free energies. Early experimental data already exist for monovalent chloride salts obtained through pioneering neutron diffraction experiments employing the isotopic substitution technique, albeit with limited resolution and momentum transfer (q) range. Modern X-ray diffraction studies on the simple chlo-
ride melts are scarce, and old data were typically obtained from standard X-ray sources with a narrow q-range, causing uncertainties in the positions and heights of peaks in the corresponding PDF due to truncation effects. In our current study, we overcome these obstacles by utilizing high-energy X-rays (λ = 0.1667 Å, 74.4 keV) to measure structure functions, S(q), with an extended q-range and low statistical noise, which by Fourier transformation provide more accurate PDFs.

One of the critical objectives of molten salt-based technologies in electrochemical applications is to attain high ionic conductivity. Trends in diffusivities and conductivities will be contrasted with available experimental data in the literature allowing us not only to test models but also to establish possible connections with rate theories of ion exchange. Several computational studies on molten alkali halide salts and their mixtures at different concentrations and temperatures already exist in the literature, but questions still remain about transport and ion exchange dynamics on a microscopic scale. For example, it has been established that a small cation (such as Li+) in neat molten salts has high internal mobility, which is reduced significantly when mixed into a salt of a larger cation (such as K+) at a low molar fraction of the smaller cation. Although this finding about the cationic dynamics is remarkable, rates and mechanisms describing the dynamics of anion-exchange around cations remain unknown even in the case of the neat salts. We will attempt to obtain such mechanistic information and reveal the effect of barriers, mass, volume, and the coupling to solvent by applying and in cases extending the formalism of Marcus theory and transition state theory.

2 Materials and Experiments

2.1 Molten salt samples

The anhydrous monovalent chloride salts LiCl, NaCl, and KCl (99.9% purity) were purchased from a commercial supplier. Approximately 50 grams of each salt were loaded into quartz tubes in a glove box under a nitrogen atmosphere. Each tube was fitted with a compression fitting connected to a Teflon stopcock, taken in a glove box under a nitrogen atmosphere. Each tube was fitted approximately 50 grams of each salt were loaded into quartz tubes (99.9% purity) were purchased from a commercial supplier. The anhydrous monovalent chloride salts LiCl, NaCl, and KCl were added to quartz tubing allowed for the attachment of a compression fitting so that vacuum could be pulled to flame seal the capillaries, once they were loaded with samples (Fig. S1 in the ESI). After the extra tubing was fitted to each capillary, they were taken into the glove box. Each salt was crushed into a fine powder using a mortar and pestle and then the additional tubing added to each capillary was used as a funnel, and the powder was added until the capillary was full. After the capillary was packed with the appropriate amount of salt, it was taken out of the glove box with a compression fitting and flame sealed under a 1 × 10−3 Torr vacuum.

2.2 X-ray scattering measurements

Total X-ray scattering experiments were performed at the National Synchrotron Light Source II (NSLS-II), 28-ID-1 beamline, using X-ray photons of wavelength 0.1667 Å and a beam of cross-sectional area 0.25 × 0.25 mm2, giving an accessible q-range up to 28 Å−1. The complete experiment comprised the measurement of the X-ray diffraction patterns for a molten salt in a cylindrical quartz capillary and the empty capillary. Data were collected using a customized furnace (details are reported in our previous work on molten salts) to hold a quartz capillary at high temperature (958 K (LiCl), 1148 K (NaCl), and 1173 K (KCl)). These temperatures were explicitly selected to compare our new results with previously reported X-ray/neutron scattering experiments. Calibration of the precise sample to detector distance (207.68 mm), detector tilt and rotation, and beam center was performed using a Ni powder as the standard. The measured background intensity was subtracted, and standard corrections were made for inelastic Compton scattering and sample self-attenuation using the PDFgetX2 software. The obtained data were subsequently normalized to the average electron density given by the weighted sum of the atomic X-ray form factors, yielding the total structure function S(q):

\[ S(q) = \frac{I_{oh}(q) - \sum_{\alpha} x_{\alpha} f^{2}_{\alpha}(q)}{\left(\sum_{\alpha} x_{\alpha} f^{2}_{\alpha}(q)\right)^{2}} \]  

where I_{oh} is coherent scattering intensity, x_{\alpha} and f^{2}_{\alpha}(q) are the molar fraction and q-dependent X-ray atomic form factor of species α, respectively, and q denotes the magnitude of the scattering vector (q = (4πsinθ)/λ, where 2θ is the scattering angle, and λ is the incident X-ray wavelength). Notice that in the current article we define the structure function S(q) as in references and 70. As defined here, S(q) goes to zero at large q and corresponds to what in prior literature Keen refers to as F^s(q), "the X-ray weighted total-scattering structure factor". In other words, our S(q) is F^s(q), and we provide this remark here to avoid any confusion with different terminologies and definitions commonly used in total scattering.

In this study, the reduced pair-distribution function (D(r)) in the range q_{min} = 0.6 Å−1 to q_{max} = 16 Å−1 is defined as

\[ D(r) = \frac{2}{\pi} \int_{q_{min}}^{q_{max}} qS(q)\sin(qr) dq. \]  

This definition has the operational advantage that direct comparison can be made between simulations and experiments without the need to explicitly measure the number density \rho_0, which is already contained in D(r) as the slope of the function at low r. Experimental S(q) and D(r) for the entire q-range (up to 18 Å−1) and r-range (up to 20 Å) considered in our experiments are presented in Fig. S2 in the ESI. A subcomponent of D(r) is defined...
as

\[ D_{\alpha\beta}(r) = \frac{2}{\pi} \int_{q_{\text{min}}}^{q_{\text{max}}} q S_{\alpha\beta}(q) \sin(qr) dq \] 

(3)

where \( S_{\alpha\beta} \) is the partial structure function defined in references 69 and 70.

3 Simulation Protocol

To interpret our X-ray measurements and to better understand the validity of different models, we carried out three different types of MD simulations based on (1) density functional theory (i.e. \textit{ab initio} MD (AIMD)), (2) a non-polarizable ‘rigid’ ion model (RIM) 69,79 and a polarizable ion model (PIM). 69,80,81 As it will become apparent, structural information from simulations at different levels of theory shows reasonably good agreement with our experimental findings, providing confidence in the interpretation of experimental observations. Due to the prohibitive cost associated of collecting sufficient statistics, transport properties including diffusion coefficients and ionic conductivities were computed using both the PIM and RIM but not AIMD.

3.1 MD simulations using the RIM and PIM

To study the structure and dynamics of LiCl, NaCl, and KCl, simulation boxes containing 864 ion pairs were generated. Initial configurations were those of the crystal structures of the salts at the experimental liquid densities. For convenience, all systems were first equilibrated using the RIM. 25,29 Crystals were first melted at 3000 K and then cooled to target temperature (958 K for LiCl, 1148 K for NaCl, and 1173 K for KCl) over 5 ns at constant volume. After this, systems were further equilibrated in the constant pressure and temperature (NPT) ensemble for another 3 ns (at 1 atm and target temperature). The last 1 ns of these runs was used to compute \( S(q) \) (see reference 69 for details), as well as \( D(r) \) and its subcomponents. RIM simulations were performed using the LAMMPS package 82 with a time step of 1 fs.

Final snapshots from the RIM simulations were taken as initial points for simulations using the PIM. 80,81 Each system was first subject to a 200 ps simulated annealing equilibration at constant pressure (at 1 bar) that increased their temperature to 1640 K and cooled back to target temperatures. Further equilibration and production runs in the NPT ensemble (at 1 bar and target \( K \) and cooled back to target temperatures. Further equilibration and production runs in the NPT ensemble (at 1 bar and target temperature) as well as the calculation of \( S(q) \) and \( D(r) \) followed the same scheme as described for the RIM simulations. PIM simulations were performed using the CP2K software 25 where trajectories were run in the NVT ensemble using the Nosé-Hoover thermostat with a 1 ps time constant. 84,85,86 Valence electrons were treated explicitly at the DFT level employing the revPBE functional 89 and the DZVP-MOLOPT basis set 90 with density from a cutoff of 400 Ry. The core electrons on all atoms were represented by revPBE pseudopotentials, and we used the Grimme’s D3 long-range dispersion correction to the DFT functional. 91 A time step of 0.5 fs was used to generate a 120 ps trajectory and the last 100 ps of that trajectory were used for computing \( S(q) \) and \( D(r) \).

3.3 Calculation of transport properties

Diffusion coefficients were calculated from 1 ns RIM and PIM simulations at 1100 K in the NVT ensemble with density selected as the average resulting from NPT equilibration as described in subsection 3.1 using the expression

\[ D = \frac{1}{6} \lim_{t \to \infty} \frac{\partial}{\partial t} \langle (r_i(t) - r_i(0))^2 \rangle, \] 

(4)

where \( (r_i(t) - r_i(0))^2 \) represents the squared displacement of ion \( i \) at time \( t \), and the \( \langle ... \rangle \) indicates an ensemble average for all ions of the same ion type. Since the diffusion coefficient is system-size dependent, to get a best estimate to its large size limit we applied the Yeh-Hummer correction 82 as defined in Eq. 5

\[ D_{\text{iso}} = D(L) + 2.837298 \frac{k_B T}{6\pi \eta L} \] 

(5)

In Eq. 5, \( D_{\text{iso}} \) is the infinite system size diffusion coefficient, \( D(L) \) is the calculated diffusion coefficient from a cubic box with edge length \( L \), and \( k_B \), \( T \), \( \eta \) are the Boltzmann constant, temperature and shear viscosity respectively. For the different models, \( \eta \) values used in Eq. 5 were computed using a protocol similar to that described in reference 92.

To compute the ionic conductivity of the RIM and PIM, 30 independent trajectories were simulated using the same NVT simulation protocol and data were saved every 10 fs for analysis. These trajectories were 1 ns in duration in the case of the RIM and 200 ps in the case of the PIM. Data at 100 fs intervals from these 30 independent trajectories were also used to evaluate standard de-
Ionic conductivity was computed using the Green-Kubo formula:

$$\sigma = \frac{1}{3V_{kB}T} \int_0^\infty (J(t) \cdot J(0)) dt,$$  \hspace{1cm} (6)

where,

$$J(t) = \sum_{i=1}^N q_i v_i(t).$$  \hspace{1cm} (7)

Here, $J(t)$ is the charge flux, $q_i$ is the formal charge of each ion, and $v_i(t)$ is the velocity of each ion. The ionic conductivities were computed from the average of the 30 independent trajectories. To estimate errors in our conductivity calculations, a bootstrapping method was utilized in which 30 trajectories were randomly selected 50 times, (a given trajectory can be selected multiple times) and the standard deviation was computed from the result of these 50 trials. We also report the computed molar conductivity as defined in Eq. 8

$$\sigma_M = \frac{2N_A}{3N_{kB}T} \int_0^\infty (J(t) \cdot J(0)) dt$$ \hspace{1cm} (8)

Here, $N_A$ and $N$ are the Avogadro number and the total number of ions, respectively.

### 3.4 Rate Theories for Anion Exchange

In an anion exchange process around a cation, the equilibrium coordination structure of the cation before the exchange is the same as its equilibrium coordination structure after the exchange. However, since we want to develop a Marcus-based theory for our systems we need reactants and products with coordination states that are distinct. Such distinction can be artificially introduced by theoretically treating a selected cation-anion pair as the solute and the rest of the ions in the system as part of the solvent. Of course, any neighboring anion can be considered as the selected counter-ion and this is effectively what we do upon averaging. For the purpose of developing our model, we consider a 2D-reaction coordinate $R_e = (r,n)$, involving (1) the cation-anion distance ($r$) of the selected ion pair and (2) the number of remaining anions ($n$) from the solvent coordinating with the cation of this pair. For this reaction coordinate, reactant, product and transition state are pictorially depicted in Fig. 1 as concentric, alternatively arranged, anionic and cationic solvation shells around the cation $X^+$. In Fig. 1 the selected Cl$^-$ (labeled red) that forms a hypothetical solute ion pair with $X^+$ in the reactant state, exchanges with a "solvating" Cl$^-$ (labeled green) from the outer anionic solvation shell leading to the product state. For the alkali chloride salts, our goal is to establish the trend in time-scales for this exchange, as well as to better understand the key contributing factors (free energy barriers, solute-solvent couplings, molar volume, etc.), that lead to said trend. The anion exchange process occurs via solvent rearrangements that drive the solvent Cl$^-$ from the outer anionic solvation shell to the boundary of the inner anionic solvation shell of $X^+$.

This overcrowded anionic environment for $X^+$ (i.e., the transition state) eventually evolves into a product state where our original solute ion pair has separated and the solvent Cl$^-$ now coordinates with $X^+$. When the solute Cl$^-$ is far away from $X^+$, the latter is fully solvent-exposed and has the actual equilibrium coordination number ($n_{ave} = n$) hypothetically set to 6 in Fig. 1 whereas $n \neq n_{ave}$ in the reactant state.

For our 2D reaction coordinate, we compute the joint probability distribution $(P(r,n))$ of $r$ and $n$ from which a 2D-free energy landscape can be obtained through the expression

$$W(r,n) = -k_B T \ln P(r,n).$$

We also define the quantity $dN = 4\pi r^2 \rho_A P(r,n) dr$ as the number of cases for which the solute $X^+$-Cl$^-$ distance is between $r$ and $r+dr$ and the $X^+$-solute Cl$^-$ coordination number is between $n$ and $n+dn$ ($\rho_A$ is the anionic number density). As required in our rate theory formalism, we define $n$ in terms of a smooth function (continuously differentiable), $f(r_i)$:

$$n = \sum_{i=1}^{N_{mol}} \frac{1 - (r_i/r_i^*)^{12}}{1 - (r_i/r_i^*)^{24}} = \sum_{i=1}^{N_{mol}} f(r_i),$$  \hspace{1cm} (9)

where $r_i$ is the distance between $X^+$ and the $i^{th}$ Cl$^-$ out of the $N_{mol}$ solvent anions; $r_i^*$ is the location of the boundary of the first solvation shell defined from the cation-anion $g_{a\beta}(r)$ shown in Fig. S3 in the ESI.

According to Marcus theory, an anion exchange process can be described by considering the reactant (R) and product (P) states as 1D parabolic functions of $n$, which are extracted from $W(r,n)$ as follows:

$$W_R(n) = \frac{1}{2} K_R (n-n_R)^2$$

$$W_P(n) = \frac{1}{2} K_P (n-n_P)^2 + \Delta W.$$  \hspace{1cm} (10)

In equation 10 $K_R$ and $K_P$ are the reactant and product parabola curvatures with minima at $n = n_R$ and $n = n_P$, respectively. $\Delta W$ is the free energy difference between the product and reactant equilibrium states ($W_P(n_P) - W_R(n_R)$). To obtain $W_R(n)$ and $W_P(n)$, we extract parabolically fitted slices through $W(r,n)$ at $r = r_N$ and $r = r_P$ that represent the equilibrium cation-anion separations in reactant and product states, respectively. In our simulations, we find that $r_N$ and $r_P$ derived from the minima in $W(r,n)$ are identical to those derived from the 1D free energy, $W(r) = -k_B T \ln (g_{a\beta}(r))$. The extracted parabolas are diabatic states and cross at a particular coordination number that defines the transition state ($n^*$), representing the aforementioned anionically overcrowded state. The Marcus parabolas can couple in their overlapping region, resulting in a lower and a higher adiabatic free energy surface. As we will see in the results section, in all our cases, reactant and product minima are well-separated and the parabolas cross in the “normal region” (opposite sides). Therefore, reactant-to-product transitions can be described by the transition dynamics on the lower adiabatic free energy surface.

The free energy barrier ($\Delta W^*$) for the reactant-to-product transition takes a simple expression (Eq. 11) when $K_R = K_P$. In equation 11 $\Delta W^*$ depends on $\Delta W$ and the solvent reorganization energy ($\lambda = W_R(n_P) - W_R(n_R)$) which is the energy cost required for changing the equilibrium reactant coordination state to the equi-
librium product coordination state.

\[
\Delta W^\dagger = \frac{(\lambda + \Delta W)^2}{4\lambda}.
\]  

(11)

Instead, when \(K_R \neq K_P\), the barrier is sensitive to the difference between these curvatures, and there are two crossing points. The reactant-to-product transition should most-likely occur through the crossing point with the lowest barrier given by:

\[
\Delta W^\dagger = \frac{K_P(K_R + K_P)}{\Delta K^2} \lambda + \frac{K_R K_P}{\Delta K} \Delta W

- \sqrt{2(n_R - n_K)K_R K_P} \Delta W \left[ K_P \lambda + \Delta K \Delta W \right].
\]  

(12)

Using the free energy barrier derived from Marcus theory, Eq. \[13\] provides a simple approximation to the transition rate between reactant and product states based on Wigner’s transition state theory (TST) \[13,24,25\]:

\[
k_M = \frac{k_B T}{h} \exp\left[ -\Delta W^\dagger / k_B T \right],
\]  

(13)

where \(h\) is the Planck constant. In equation \[13\], the product of a prefactor, \(k_B T / h\), and a Boltzmann factor associated with the free energy barrier account for the equilibrium flux of trajectories through the dividing surface (transition state) between the reactant and product states. However, Schenter et al. \[25\] showed that an accurate determination of the equilibrium flux for a particular reaction coordinate generally requires the determination of the prefactor specific to that reaction coordinate. Because of this, in our calculations, we will use the recently-extended TST formulation by Roy et al. \[24\], where the transition rate between different states depends on the prefactor, the free energy barrier, and the transmission coefficient (representing barrier-recrossing) all of which are sensitive to the choice of the reaction coordinate. Specifically, our first estimation of anionic exchange rates will combine Marcus theory’s lower adiabatic free energy surface, \(W(n)\), with Roy’s TST formalism for \(n\) as given in Eq. \[14\]:

\[
k_{TST}^n = \frac{Z_n}{2\pi \beta} \exp\left[ -\beta W(n^\dagger) \right].
\]  

(14)

Here, \(n^\dagger\) is the location of the transition state (barrier) on the lower adiabatic surface, which is the same as the location of the crossing point between the two Marcus diabats, \(\beta = 1/k_B T\), \(Z_n = \Lambda/\mu\), \(\mu\) is the reduced mass of a select cation-anion pair, and \(\Lambda\) is given by:

\[
\Lambda = \left[ \sum_{i=1}^{N_{an}} (f_i')^2 + \sum_{j=1}^{N_{sol}} \sum_{i=1}^{N_{an}} N_{sol} (f_i f_j') (\hat{r}_i \cdot \hat{r}_j) \right]^{1/2},
\]  

(15)

where \(m_{cat}\) is the mass of the cation. The derivative, \(f_j' = \frac{df_j}{dr_j}\), is obtained from the function, \(f_j(r_j)\), (defined in Eq. \[9\]) with respect to the distance between the cation and the \(j^{th}\) solvent anion, as given in Eq. \[16\]:

\[
f_j' = \left( \frac{a}{1 - (\frac{Z_a}{Z})^{2a-1}} \right) \left( \frac{a}{1 - (\frac{Z_a}{Z})^{2a-1}} \right)^{3a-1}.
\]  

(16)

In Eq. \[16\], \(a = 12\), and we will see in the results section that for each alkali halide salt, \(Z_a\) in Eq. \[14\] is an essentially conserved quantity that minimally fluctuates over time around its average value. Another key result will be that whereas there is no particular trend for \(\Lambda\) across the systems, there is a clear trend for \(Z_a\) that is dominated by the value of \(\mu\). We highlight that in Roy’s formulation \[24\] the dynamics associated with an arbitrary reaction coordinate has an associated mass-like quantity that enters the prefactor in \(k_{TST}\). As an example, when using \(n\) as a reaction coordinate the corresponding mass-like quantity in equation \[14\] is \(1/Z_n\), with \(Z_n = \Lambda/\mu\); instead if the reaction coordinate is defined as the interionic distance between a cation and an anion (vide infra) the mass-like quantity is \(\mu\) (i.e., \(Z_a = 1/\mu\)). We will see in the results section that independent of the reaction coordinate
If Marcus theory was exact, upon arrival at the crossing point of the reactant and product parabolas from the reactant equilibrium coordination state, an anion spontaneously dissociates from the paired state with the cation leading to the product coordination state. However, it is possible that there is a non-vanishing barrier along the ion pair distance at the crossing point \( (\Delta W_{n}^t) \), making the dissociation event nonspontaneous. \( \Delta W_{n}^t \) can be determined by extracting a slice from \( W(r,n) \) at \( n = n^t \); the first barrier on that slice provides \( \Delta W_{n}^t \). If \( \Delta W_{n}^t \gg 0 \), it must be accounted for as an additional barrier for the reactant-to-product transition. Furthermore, rapid fluctuation of the surrounding solvent coupled to the motion along the coordination number can cause recrossing at the crossing point, resulting in effectively slower transition rates. Such nonequilibrium solvent effects can be treated by utilizing the semiclassical approach of Landau and Zener, which corrects Marcus rates through the determination of the transmission coefficient \( k_{LZ} \). \( k_{LZ} \) is defined in terms of the probability \( P \) of reactive transitions through the crossing region and the location of the crossing region.

\[
 k_{LZ} = \begin{cases} 
 2P/(P+1) & \text{“normal region” (opposite sides)} \\
 2P(1-P) & \text{“inverted region” (same sides)} 
\end{cases} 
\]  

Since for all the molten alkali chloride studied here the Marcus parabolas cross on opposite sides, only the “normal region” expression in eq. (17) is relevant to our study. \( P \) depends on the coupling \( C \) between the reactant and product parabolas and the positive traversal velocity \( (v_n) \) in coordination number space at the crossing point:

\[
P = 1 - \exp\left(-\frac{2\pi c^2}{\hbar v_n[S_2-S_1]}\right). 
\]  

where \( \hbar = h/2\pi, S_{1,2} = \frac{dW(n)}{dn}\big|_{n=n^t} \) are the slopes of the parabolas at the crossing point, and \( v_n \) is the mean value of the traversal velocity distribution, \( C(v) \), at the crossing point obtained from the phase space trajectories of the coordination number. Our simulations showed that the velocity distribution at the crossing point has a Gaussian form: \( C(v) = C_0 \exp(-v^2/\sigma_n^2) \), and therefore \( v_n \) can be obtained as \( v_n = \int v C(v) dv \). \( C \) is evaluated at \( n = n^t \) using the formula, \( C(n) = \frac{\kappa_{LZ} + \kappa_{RF}}{\kappa_{TC}} \sqrt{\frac{W_r(n)}{W_r(0)}} \left| \frac{W_p(n)}{W_p(n_p)} \right| \), where the prefactor disappears if the reactant and product parabolas have equal curvatures. Note that, \( P \) is mostly dominated by the reactant-product coupling strength, \( C \), affecting the transition rates. A stronger reactant-product coupling leads to a larger probability of reactive transition resulting in a larger transmission coefficient and a faster transition rate for two distinct Marcus parabolas that cross at the “normal region”. Now, after incorporating the additional, non-vanishing barrier at the transition state \( (\Delta W_{n}^t) \), the transmission coefficient \( (k_{LZ}) \), and \( k_{M} = k_{TST} \), the corrected Marcus rate expression \( k_{M}^{Correct} \) takes the form of:

\[
k_{M}^{Correct} = k_{LZ} \sqrt{\frac{Z_T}{2\pi \beta}} \exp\left[-\frac{\beta (W(n^t) + \Delta W_{n}^t)}{2\beta \pi \hbar^2} \right] 
\]  

At this point, we recognize \( (1/\sqrt{Z_T}) v_{R} = (1/\sqrt{Z_m}) \int_{n^t}^{n^t} d\epsilon \exp\left[-\beta W(n)\right] \) as a type of mass-weighted configuration space region in coordination number which we dub “mass-weighted reactant volume” since the integral sums over the Boltzmann-weighted volume elements of coordination number \( (dn) \) in the reactant region (as gleaned from the upper limit of the integral, which is the location of the transition state, \( n^t \)). It will become apparent in the results section that \( (1/\sqrt{Z_{mR}}) v_{R} \) plays an important role in distinguishing cationic size effects on the anion exchange rates.

A complementary analysis to the one discussed above in the case of \( n \) as a reaction coordinate can be carried out instead utilizing TST with the distance \( r \) as the reaction coordinate. This approach has been extensively used to study ion-pairing and solvent exchange kinetics\,[18,19,20,50–54] where nonequilibrium solvent effects are treated by computing the transmission coefficient with a variety of methods such as that of Kramers, the Grote-Hynes’s stable state picture, or the reactive flux (RF) method by Chandler and Bennett. We see from Eq. (20) corresponding to Roy’s approach,\,[14] that the expression is also influenced by a mass-weighted configuration space “reactant volume”, \( (1/\sqrt{Z_T}) v_{RF} = \sqrt{\beta} v_{RF} = \sqrt{\beta} \int_{r^t}^{r^t} 4\pi r^2 d\epsilon \exp\left[-\beta W(r)\right] dr \), in addition to the barrier \( (W(r^t)) \) and the transmission coefficient \( (k_{RF}) \).

\[
k_{Correct}^{TST} = k_{RF} \sqrt{\frac{Z_T}{2\pi \beta}} \frac{4\pi r^t \exp\left[-\beta W(r^t)\right]}{\int_{r^t}^{r^t} 4\pi r^2 \exp\left[-\beta W(r)\right] dr} 
\]  

Simple manipulations of the above expression result in

\[
k_{Correct}^{TST} = k_{RF} \sqrt{\frac{Z_T}{2\pi \beta}} \frac{4\pi v_r r^t \exp\left[-\beta W(r^t)\right]}{\int_{r^t}^{r^t} 4\pi r^2 v_{RF} (r) \rho_\lambda dr} 
\]  

As a result, the averaged reaction coordinate (\( \bar{v}_{r} = \sqrt{\beta} v_{RF}/\mu \)) is a quantity proportional to the average rel-
4.1 Structure of molten alkali chloride salts from X-ray scattering measurements

High-energy X-ray data were collected at Brookhaven National Laboratory (NSLS-II) using the methodology described in subsection 2.2. These measurements provide $S(q)$ as defined in Eq. 1 and via the transformation defined in Eq. 2, the reduced pair distribution function $D(r)$. We use both of these functions to interpret liquid structure and provide a benchmark comparison with simulation results. A key aspect in the generation of $D(r)$ is the finite maximum $q$ value ($q_{\text{max}}$) of the measured $S(q)$. Early experiments had a short cutoff value that likely introduced significant truncation errors limiting the accuracy of $D(r)$ particularly in key short-range regions. These issues are minimized by measuring $S(q)$ up to a sufficiently high momentum transfer at which $S(q) \to 0$. Unlike standard laboratory X-ray instruments, the high energy photons from modern synchrotron sources achieve a wide $q$-range at a relatively small scattering angle, significantly minimizing truncation issues and also reducing sample attenuation, multiple scattering, and other angle-dependent correction factors.\cite{Takagi2004, Takagi2005, Takagi2006} Fig. 2 nicely illustrates the effect that a small $q_{\text{max}}$ in $S(q)$ (and hence in the upper limit of the integral in Eq. 2) has on $D(r)$ by comparing the old X-ray diffraction results by Takagi et al.\cite{Takagi2004} with our experiments for molten KCl at 1173 K. As can be seen from the top panel, there are still noticeable oscillations in $S(q)$ beyond $\sim 7$ Å$^{-1}$ ($q_{\text{max}}$ achieved by Takagi) and a more reasonable truncation value should be $q \sim 12$ Å$^{-1}$. As expected, $D(r)$ in the bottom panel of Fig. 2 shows significant differences at all ranges of $r$ across experiments. These are mostly due to truncation differences since on the range measured by Takagi and coworkers the actual $S(q)$ is very similar to that measured at NSLS-II. This example provides an important cautionary tale as many of the pioneering X-ray and neutron scattering experiments on molten salts (from which quantities such as coordination numbers and first neighbor solvation geometries are based) were limited due to technical constraints to small $q_{\text{max}}$ values.

To examine whether our MD simulations can accurately describe the structure of the molten chloride salts, we computed X-ray structure functions, $S(q)$, at different levels of theory to compare with those from the synchrotron X-ray scattering experiments. Figs. 6-8c show that across systems, simulations using each of three models (PIM, RIM, and ab initio) capture features of the experimental $S(q)$ fairly well, but the PIM model appears to be overall slightly more accurate when comparing peak positions and intensities. This is in part because the chloride ion is significantly polarizable, because AIMD results may be sensitive to the chosen DFT flavor, and also because DFT simulations boxes are necessarily smaller than those used for the PIM and RIM simulations.

The interpretation of peaks in $S(q)$ can be achieved by decom-

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Footnote 1: We believe a typo has been introduced in expression \( \frac{\sum m K_m}{\sum m} \) of Eqns. 7 and 10 from reference 24 where a factor of $N$ appears to be missing. Eq. 7 also appears to be missing a factor of $r$. To the best of our understanding, to convert the reduced intensity data, $n(x)$, in Fig. 2 of the aforementioned article to our notation as shown in Fig. 2 (top) we must use $S(q) = \frac{x \sum m}{\sum m K_m}$. In the case of Fig. 2 (bottom) we use the definition of $g^*(r)$ provided in eq. 9 of reference 24.
posing the function into contributions from the different ion-pair subcomponents (the partial structure functions, $S_{\alpha\beta}(q)$). Figs. 3d-f show how cation-cation, anion-anion, and cation-anion correlations contribute to the total $S(q)$ determined from the PIM simulations. The inherent structural characteristic of molten salts is charge alternation, and its signature is a positive-going peak arising from the contributions of same-charge ions at about the same $q$ value as a negative-going peak (also known as an antipeak) resulting from spatial correlations between opposite-charge ions. For LiCl, these charge-alternation peaks and the concomitant antipeak are present around $\sim 2$ Å$^{-1}$; peaks and antipeaks corresponding to this feature move to lower $q$ values ($\sim 1.75$ Å$^{-1}$ for NaCl and $\sim 1.6$ Å$^{-1}$ for KCl) with larger cation size. This is because these peaks and the corresponding antipeak are linked to the distance between ions of the same charge alternated by ions of opposite charge which becomes larger as the cation size increases. Summing the three contributions (cation-cation, anion-anion and cation-anion) results in a prominent peak in the total $S(q)$ for LiCl but the peak diminishes in intensity for NaCl and is completely absent for KCl. This is purely related to lack of contrast in X-ray experiments; the X-ray form factor for Li is small but that for Cl is large and the sum of intensities of the two peaks (Li-Li and Cl-Cl) are not cancelled by that of the Li-Cl antipeak at the same $q$ value. Instead, there is essentially complete cancellation of peaks and antipeaks for KCl resulting in a missing charge alternation feature in the overall $S(q)$. As is obvious from the two large (and nearly identical) peaks and the antipeak in Fig. 3d, this does not mean that there is no charge alternation in KCl but instead, that there is a poor contrast in the technique. For example, the result would be different if one was to use neutron weighting factors instead of those for X-ray. The peak at a larger $q$ value than charge alternation is what we commonly call the "adjacency peak". In general, this peak is associated with short-range structural interactions between nearest neighbors. In the case of salts, these are commonly opposite-charge ion interactions. This can be clearly gleaned from the similarity between the total $S(q)$ and the cation-anion partial $S(q)$ above $\sim 2$ Å$^{-1}$ in Figs. 3e-f. The case of LiCl is particularly interesting since there is a major weight difference in the X-ray form factors between the cation and the anion. This results in no apparent Li-Cl adjacency peak; in Fig. 3d, the black line does not look like the blue line between 2.5 and 3 Å$^{-1}$.

Real space $D(r)$ functions obtained by the Fourier transform of the total structure functions (Eq. 2) are depicted in Figs. 4a-c, and compositionally resolved partial $D_{\alpha\beta}(r)$ from the PIM simulations (as defined in Eq. 3) in Figs. 4d-f. Corresponding pair distribution functions $S_{\alpha\beta}(r)$ are provided in Fig. S3 of the ESI. These figures show that the shortest contact ion pair distance (dominating the first peak in $D(r)$) is achieved for LiCl, followed by NaCl and KCl as is obviously expected from the trend in cationic sizes. It is worth noting that the reduced pair distribution function, $D(r)$, emphasizes the large-$r$ portion of the data and this helps highlight ionic correlations beyond short-range interactions. Partial $D_{\alpha\beta}(r)$ from our MD simulations facilitate the interpretation of real-space data by showing how particular ion-ion correlations are manifested in the total $D(r)$. As the number of electrons in the cation increases, so does the cation-cation contribution to $D(r)$. Analogous to the case of the partial subcomponents of $S(q)$ discussed in the previous paragraph, this is simply an issue of contrast in the X-ray technique. Overall, our results appear to indicate that simulations, particularly the less expensive classical ones, can be used for the interpretation of experimental results.

4.2 Dynamical Macroscopic Properties

In this section, we present results on diffusivity and conductivity and discuss these in the context of several prior studies in the
Fig. 4 Comparison between our simulated and experimental $D(r)$ functions for LiCl at 958 K, NaCl at 1148 K, and KCl at 1173 K (top row) and interpretation of different peaks in terms of the sub-components of the total $D(r)$ obtained from the PIM simulations (bottom row). (See also pair distribution functions $g_{\alpha\beta}(r)$ in Fig. S3 of the ESI.)

Fig. 5 Variation in the diffusion coefficient, molar conductivity, and conductivity with increasing ionic radius of the cation obtained from experiments\cite{119,120} as well as the PIM and RIM for LiCl, NaCl, and KCl at 1100K.

(a) depicts the self-diffusion coefficients and ionic conductivities computed from the PIM and RIM simulations at 1100K (see Eq. 4, 5, 6, and 7) in comparison to experimental data\cite{119,120} as a function of cationic size. Fig. 5(a) shows that whereas both models correctly capture the order of magnitude and trends in diffusion coefficients, there are clear differences between these. Experimental and simulation results indicate that there is a marked distinction between the diffusivity of Li$^+$ which is high and that of Na$^+$ and K$^+$ which are lower and quite similar. Fig. 5(c) shows experimental and computational conductivities (see Table S1 in the ESI for values and error estimations). Here too there are differences across models and experiments but the trend is clear; there is a decrease in conductivity with an increase in cationic size. The reader is reminded that there is an explicit dependence on the volume in the conductivity (Eq. 6) and since the molar volume for each of these salts is different, trends in Fig. 5(c) must concomitantly capture the differences in molar conductivity and molar volume. We notice that trends in the molar conductivity (Fig. 5(b)) are much more similar to those in the diffusion coefficients, where going from NaCl to KCl changes are much more modest than going from LiCl to NaCl. We find that neither model perfectly captures the experimental trend in molar conductivity (both models show essentially flat behavior for NaCl and KCl) but models and experiments coincide in that the change in going from LiCl to NaCl is much more pronounced (about 2.65 times larger for the experiment) than in going from NaCl to KCl.

The ionic conductivity, is made of contributions from the collective dynamics of same- and different-charge species that goes beyond the self-diffusion dynamics of cations or anions. This can
be gleaned from Eq. 6 which can be expanded as

\[ \sigma = \sum_{i=1}^{i=n} w_i D_i \]

\[ \sigma = \rho_0 v^2 / k_B T \left[ x_{cat}^2 D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C + x_{cat}^2 Z_{cat} D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C \right] \]

\[ \sigma = \rho_0 v^2 / k_B T \left[ x_{cat}^2 D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C + x_{cat}^2 Z_{cat} D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C \right] \]

\[ \sigma = \rho_0 v^2 / k_B T \left[ x_{cat}^2 D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C + x_{cat}^2 Z_{cat} D_{cat}^C + x_{an}^2 Z_{an} D_{an}^C \right] \]

In eq. 23, \( \epsilon \) is the protonic charge, \( N_{\text{cat}} \) and \( N_{\text{an}} \) are the number of cations and anions, respectively, providing the total number of ions \( N = N_{\text{cat}} + N_{\text{an}} \). \( Z_{\text{cat}} = 1 \) and \( Z_{\text{an}} = -1 \) are their charge numbers and \( x_{\text{cat}} = 0.5 \) and \( x_{\text{an}} = 0.5 \) are their mole fractions. In Eq. 23, the integrals (including the factor 1/3) represent different types of diffusion coefficients; cation-cation and anion-anion self-diffusion coefficients \( (D_{cat}^C \) and \( D_{an}^C \) \) and distinct diffusion coefficients \( (D_{cat}^C \) and \( D_{an}^C \)) and the distinct cation-anion diffusion coefficient \( (D_{cat-an}) \). Eq. 23 highlights that ionic conductivity is proportional to the weighted sum of these five different diffusion coefficients \( (D_i) \), where their prefactors represent the weights \( (w_i) \). Kashyap et al. examined the behavior of these different velocity auto- and cross-correlation functions and found that short-time ionic motion is very important to the overall ionic conductivity; they also explained that the relative motion of opposite-charge ions enhances conductivity.

Fig. 6 shows all correlation functions \( (C_r) \) as well as the weighted contributions of the different diffusion coefficients to the conductivity. As can be seen from Fig. 6, for the most part, all \( C_r \) functions decay to zero within a fraction of a picosecond (although small oscillations still persist beyond this time that contribute to their time integrals) confirming the idea that what happens at a very short time is extremely important to the overall value of the conductivity in these systems. A few things can be learned from the bottom panel in Fig. 6. First, as expected from AB-type ionic systems where there is no third neutral solvent component to act as a "momentum buffer," the cation-anion contribution to the conductivity is positive. This is in contrast to what happens to ions in solution where ion-pairing reduces conductivity. Second, the behavior of LiCl across the different weighted \( D_i \) functions is very different from that of NaCl and KCl which are much more similar. This is not to say that the \( C_r \) functions are similar for NaCl and KCl, but in most cases, the long-time weighted integrals are when compared to those for LiCl. This is particularly clear in the case of the cationic self-contribution, but is also apparent for the anionic self and distinct contributions. In the case of LiCl the positive contribution of the anionic self term is roughly cancelled by that of its distinct contribution; since cation-anion and cation distinct contributions across the three salts are not that different, one can fairly state that it is the self diffusion component for \( \text{Li}^+ \) that makes the conductivity for LiCl distinctly different from that of NaCl and KCl. Notice that distinct cation and anion contributions in Fig. 6 and Fig. 6 across the three salts appear to be in reverse order and in the case of NaCl and KCl these differences roughly cancel each other resulting in overall conductivities for NaCl larger than for KCl that follow the trends in Fig. 3-h. It is now easier to see why the drop in conductivity is much larger when going from LiCl to NaCl than from NaCl to KCl. The dominating effect is simply the high self diffusivity of \( \text{Li}^+ \) and the similarity in trends between Fig. 5 and Fig. 5 confirms this.

4.3 Rate Theories of Anion Exchange for the Alkali Halides

To motivate this section and our interest in understanding solvation-shell dynamics and ion exchange processes for the alkali halide molten salts, we remind the reader that computational work by Zhang et al. investigated around 30 ionic liquids at various temperatures, and found that, independent of the choice of system and temperature, there was a nearly linear relation between the lifetime of the closest cation-anion neighbors and both their self-diffusion coefficient and ideal or Nernst-Einstein ionic conductivity. It is therefore reasonable to conjecture that for these molten salts, counter-ion exchange could perhaps also be taken as a proxy to understand transport properties. For this purpose, we use three different techniques that provide alternative but complementary views of the anion exchange process. The first technique is based on a combination of Marcus theory and transition state theory where the reaction coordinate is the coordination number \( n \) of an alkali chloride ion pair as discussed in the methods section, the second is the more traditional transition state theory where the distance between a cation and an anion is taken as the reaction coordinate, and the third one is based on the survival probability correlation function. The first two methods require knowledge of the free energies, \( W(r) \), \( W(r,n) \), and \( W(n) \), which are presented in Fig. 7-a-g for the PIM simulations at 1100 K. Fig. 5 in the ESI provides a comparison between the PIM and RIM results, indicating a good agreement between them. The reader is reminded that \( n \) in the y-axis of \( W(r,n) \) is defined as the number of counter-anions to a central alkali metal ion not counting the special anion defined to be its pair (the progression from reactant to transition and finally product states as well as the definition of \( n \) are depicted in Fig. 7). This theoretical construct is used so that \( n \) is different (smaller) in the reactant state than in the product state even though the
The actual number of Cl\textsuperscript{−} ions surrounding the cation is the same.

For the different salts, W(r) clearly distinguishes the equilibrium distances corresponding to the contact ion pair (r\textsubscript{R}) in the reactant state (R) and the solvent-separated ion pair (r\textsubscript{P}) in the product state (P). These equilibrium distances systematically increase going from LiCl to NaCl to KCl. However, we notice that barriers along the solute cation-anion distance are very similar (~4.6-4.8 kcal/mol) for all the salts and are significantly higher than the thermal energy at 1100K (2.18 kcal/mol). Overcoming this barrier makes ion-pair dissociation an activated process. Of course, a 1D-free energy hides the fact that barriers can be circumvented if pathways are considered in multi-dimensions such as those shown by the white arrows on the 2D-free energy landscapes in Figs. 7p-f. Physically, the white arrows highlight three mechanistic steps: First, solvent-rearrangement induces activation in coordination number (n) leading to anionic overcrowding (notice that while at the transition state n < n\textsubscript{P}, the actual number of anions surrounding the cation including the one considered as the ion pair is larger than in either of the wells); second, the overcrowding reduces the barrier along the solute cation-anion distance (r), and third the dissociation of the solute ion-pair along r takes place where the solvent anion takes the place of the solute anion as the product state. Cuts on the 2D-free energy surface at r = r\textsubscript{R} and r = r\textsubscript{P} (the diabats) in Figs. 7p-g have minima that are well separated and cross in the “normal region” (opposite sides) in the Marcus theory sense. Lower and higher adiabatic free energy curves can be derived from these as discussed in the methods section and are shown in Figs. 7h-j.

Our first method to study rates of chloride exchange across the family of alkali chlorides, is the hybrid Marcus-TST approach associated with Eq. 19. All essential parameters entering this equation in the case of the PIM and RIM are provided in Table 1 (see Table S2 in the ESI for the remaining parameters including those defining Marcus parabolas and their couplings and traversal velocities at the crossing point). τ values in Fig. 8 (labeled Marcus), which are the reciprocal of the exchange rate, show a clear trend of increasing exchange times with an increase in cation size. There are several factors contributing to this; a large contributor is what we defined in the methods section as the mass-weighted “reactant volume” in coordination number space, \[ \frac{1}{\sqrt{Z_n}} V_{R}^n = \sqrt{\frac{\mu}{\Lambda}} V_{R}^n. \] The reader should notice that for a given system, Z\textsubscript{n} = \Lambda/\mu is essentially a conserved quantity (its value as a function of time is shown in Fig. S6 of the ESI), which has complex contributions of the solvent anions incorporated in \Lambda (see Eq. 15). However, \Lambda is almost the same for all three salts (considering their error bars) as shown in Table 1. Thus, the trend in Z\textsubscript{n} is mostly dominated by \mu, therefore by the mass of the cation (as all three salts have the same anion). On the other hand,
Fig. 7 One-dimensional free energy along the solute cation-anion distances ($W(r)$) (a) and their extensions to two-dimensions ($W(r,n)$) by including the coordination number of the solute cation ($n$) with the solvent anions (evenly spaced (1 kcal/mol) contours between 0 and 10 kcal/mol) (b,d,f). White arrows on $W(r,n)$ display the Marcus pathways of ion-pair dissociation. Slices through $W(r,n)$ at $r = r_R$ and at $r = r_P$ (dotted lines) and their parabolic fits (solid lines) represent reactant and product states in coordination number space used in Marcus theory (c,e,g). Using Marcus theory, the ion pair dissociation and anion exchange mechanism around a cation in molten salts can be described as the adiabatic traversal of the crossing point of the Marcus parabolas (dashed red and blue, respectively) that couple to generate lower and higher free energy surfaces (solid red and blue, respectively) (h,i,j). These results are from PIM simulations at 1100K for LiCl, NaCl, and KCl, whereas other cases are discussed in the ESI.

$V_R^n$ does not show any specific trend (see Table 1), which leads to the conclusion that $\frac{1}{\sqrt{n}}V_R^n$ is predominantly governed by the mass of the cation. For the PIM, $\frac{1}{\sqrt{\mu}}V_R^n$ increases to $\sim 2.19$ in going from LiCl ($\sim 0.36$) to KCl ($\sim 0.79$), and since its role is multiplicative (see equation 19) so does the value of $\tau$. The second most important factor in the trend for $\tau$ is the Landau-Zener transmission coefficient which is larger for LiCl ($\kappa_{LZ} \sim 0.97 - 0.98$) than for NaCl ($\kappa_{LZ} \sim 0.91 - 0.94$) and KCl ($\kappa_{LZ} \sim 0.64 - 0.80$). For the PIM, in going from LiCl to KCl, its effect is to increase $\tau$ by about 52%. The large values of $\kappa_{LZ}$, particularly for LiCl and NaCl, are indicative of a weak coupling to the solvent bath and small nonequilibrium solvent effects for $n$. Yet, differences in $\kappa_{LZ}$ are significant in distinguishing LiCl from KCl when it comes to the anion exchange rate. The last factor that contributes to differences in the anion exchange rate is the barrier ($W(n^t) + \Delta W^e(n^t)$), which is slightly larger for KCl (4.8-4.9 kcal/mol) than for NaCl and LiCl (4.4-4.6 kcal/mol). The effect from the barrier (see the term, $e^{-\beta(W(n^t) + \Delta W^e(n^t))}$ in Table 1) in differentiating the cations is small, on the order of 9% in going from LiCl to KCl for the PIM.

The reader is reminded that the conductivity across the family of alkali chlorides in Fig. 5c shares the decreasing trend shown by the exchange rate, which is inverse to the increasing trend for $\tau$ shown in Fig. 5. Furthermore, we note that as the cationic mass (size) and concomitantly $\sqrt{\mu} V_R^n$ increase, so does the macroscopic volume across the family of molten salts at the same temperature and pressure. In going from LiCl to KCl, this increase in the overall macroscopic volume has an explicit effect on their conductivity. This can be clearly gleaned from the prefactor in Eq. 6 and the fact that conductivity and molar conductivity have different trends. The latter, which loses this prefactor, flattens out significantly as a function of cationic size, particularly when going for NaCl to KCl. It is therefore notable that an increase in cationic mass/size consistent with an increase in the macroscopic sample volume goes along with a decrease in conductivity as well as in the anion exchange rate.
Fig. 8 \( \tau \) (the reciprocal of the rate of anion exchange) as a function of ionic radius from MD simulations for LiCl, NaCl, and KCl at 1100K derived from our Marcus-TST approach based on \( n \) and the TST based on \( r \) as reaction coordinates.

Fig. 9 For the three molten salts at 1100 K, time-dependent transmission coefficients \( (\kappa_{RF}(t)) \) determined from the PIM simulations at 1100 K using the reactive flux method showing strong interionic distance-solvent coupling. The values of \( \kappa_{RF} \) used in the expression for \( \tau \) are derived from the average of \( \kappa_{RF}(t) \) in the region between 1 and 2 ps (same protocol for the RIM in Fig. S7).

The second TST approach that we used to study anion exchange rates and derive \( \tau \) (see Fig. 8 labeled TST), also follows the formulation by Roy et al.\(^\text{[54]}\) but in this case for \( r \) as the reaction coordinate. The expression for the anion exchange rate is given in Eq. [21] for which the relevant quantities are \( W(r) \) (shown in Fig. 7a)), the reactive flux transmission coefficients (shown in Fig. 8 for the PIM simulation and in Fig. S7 in the ESI for the RIM simulation), the quantity \( \tilde{v}_r \) linked to the thermal average relative velocity between the cation and the anion, the volume of the sample, and the average coordination number, \( n_{\text{ave}} \), for the different cations (all provided in Table 2).

In this case, the analysis appears to be much simpler. For all three liquids, \( K_{RF} \) is essentially the same; the structural component \[ \mu^\text{x}\exp\left[-\beta W(r)\right] \frac{r^2 n_{\text{ave}}(r)}{n_{\text{ave}}} \] would make the rate of anion exchange for KCl>NaCl>LiCl, but both the anionic number density, \( N_{an}/V \), and \( \tilde{v}_r \) contribute to the opposite trend making the anion exchange significantly faster for systems with a smaller and lighter cation. Since \( N_{an} \) is the same for all systems, the trend observed in anion exchange rates appear to be simply determined by \( \mu \) (through \( \tilde{v}_r \)), and the volumetric differences across samples. This correlates well with findings in Fig. 5 where Li has the faster diffusivity, conductivity, and molar conductivity, whereas for NaCl and KCl the larger conductivity for the former is majorly influenced by the overall sample volume.

Notice that interpretation on how the similar values across theories for \( \tau \) in Fig. 8 are arrived at can be quite different when the analysis is based on \( r \) vs. \( n \) as reaction coordinates. For example, nonequilibrium solvent effects depends on how a reaction coordinate couples with the solvent bath, and when \( n \) is the reaction coordinate, those are reflected in \( K_LZ \). Values for \( K_LZ \) hinted at a relatively weak coupling to the solvent, but the numerical value differences across salts made a very significant difference to the observed \( \tau \) values as we went from LiCl to KCl. Previous studies on aqueous electrolytes\(^\text{[23,108]}\) indicate that the distance coordinate strongly couples with the solvent resulting in very small transmission coefficients (\( \ll 1 \)). Herein, we also find the interionic distance-solvent bath coupling to be strong and affecting equally all three molten salts, since the reactive flux transmission coefficients are small and similar for all of them (~ 0.20 – 0.22).

Thus, while nonequilibrium solvent effects on the distance coordinate equally influence the anion exchange rates for all salts, the same effects on \( n \) as a reaction coordinate can slow down the rate significantly smaller amount for NaCl but a significant amount for KCl when compared to LiCl. It is therefore clear that the very nature of the nonequilibrium solvent effects strongly depends on our choice of

| Table 1 Important factors in the Marcus-TST hybrib method that uses \( n \) as the reaction coordinate. |
|----------------|--------|--------|
| \( \mu \) (g/mol) | LiCl | NaCl | KCl |
| \( \Delta W_{\text{n}}^{\text{r}} \) (kcal/mol) | 2.45 | 2.66 | 2.68 |
| \( W(n) + \Delta W_{\text{n}}^{\text{r}} \) (kcal/mol) | 4.60 | 4.63 | 4.88 |
| \exp[-\beta (W(n) + \Delta W_{\text{n}}^{\text{r}})] | 0.12 | 0.12 | 0.11 |
| \( \kappa_{RF} \) | 0.97 | 0.91 | 0.64 |
| \( \Lambda \) (nm\(^{-2}\)) | 61.93±5.05 | 63.61±3.49 | 57.63±2.97 |
| \( Z_n \) (nm\(^{-2}\)/g/mol) | 10.66±0.78 | 4.56±0.25 | 3.10±0.16 |
| \( V_R \) | 1.18 | 1.42 | 1.38 |
| \( \mu V_R \) (nm\(^{2}\)/g/mol) | 0.36±0.01 | 0.66±0.02 | 0.79±0.02 |

\( \tilde{v}_r \)
reaction coordinate and the specific rate theory method used, and therefore, it is not an unambiguous factor. The reader is reminded that, for the TST approach where \( r \) is the reaction coordinate, the barrier for anion exchange is almost identical (\( W(r^*) \approx 4.6-4.9 \text{ kcal/mol} \)) across the family of alkali halide salts. The barrier enters the structural factor \( \langle r^2 \rangle_{\text{diff}}(r^*) \) through the value of the pair distribution function at \( r^* \), and the overall contribution of this factor is opposite to the actual trend of anion exchange. In the Marcus-TST approach where \( n \) was the reaction coordinate, the total barrier (\( W(n^*) + \Delta W_r^* \)) was also quite similar (\( \sim 4.4-4.9 \text{ kcal/mol} \)) across systems. Thus, consistent across the two methods so far discussed, barrier effects do not determine the cationic size effect on the rates of anion exchange. This leads us to the overall conclusion that \( \mu \) (and therefore the mass of the cation) is an important factor in both rate theories, while solvent effects are important to the Marcus-TST approach based on \( n \), and differences in molar volume of the melt are important in the TST approach based on \( r \).

As a final and more direct way to interrogate the process of anion exchange across our systems, we utilized the method proposed by Impey, Madden, and McDonald (IMM) to compute the residence time of a solvent molecule (in our case the chlorides) in the first solvation shell of the alkali metal cations.\(^{121}\) The IMM method determines the survival probability \( (P(t,t+\delta t,r^*)) \) of a anion in the first solvation shell of a cation where \( P \) is assigned a value of 1 when an anion is found in the first solvation shell at both times \( r \) and \( t + \delta t \) and during this time interval the anion does not leave the shell for any continuous period of time greater than \( r^* \); otherwise, \( P \) is assigned a value of 0. Numerically, in the condensed phase, particles at an artificial boundary are constantly recrossing it, but this does not mean that they have left the first solvation shell. This is why a tolerance value \( r^* \) is defined to treat unsuccessful recrossing, i.e., when anions cross the boundary of the first solvation shell but rapidly return to the shell.

Table 2 Important factors in the TST method that uses \( r \) as the reaction coordinate.

<table>
<thead>
<tr>
<th></th>
<th>LiCl</th>
<th>NaCl</th>
<th>KCl</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \mu ) (g/mol)</td>
<td>5.81</td>
<td>13.95</td>
<td>18.59</td>
</tr>
<tr>
<td>( v_r ) (nm/ps)</td>
<td>1.26</td>
<td>0.81</td>
<td>0.70</td>
</tr>
<tr>
<td><strong>PIM</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( \kappa_{RF} )</td>
<td>0.20</td>
<td>0.22</td>
<td>0.22</td>
</tr>
<tr>
<td>( V ) (nm(^3))</td>
<td>42.54</td>
<td>53.89</td>
<td>71.46</td>
</tr>
<tr>
<td>( n_{ave} )</td>
<td>4.34</td>
<td>5.16</td>
<td>5.18</td>
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<tr>
<td>( r^* ) (nm)</td>
<td>0.36</td>
<td>0.41</td>
<td>0.45</td>
</tr>
<tr>
<td>( W(r^*) ) (kcal/mol)</td>
<td>4.83</td>
<td>4.65</td>
<td>4.75</td>
</tr>
<tr>
<td>( e^{-\beta W(r^*)} )</td>
<td>0.11</td>
<td>0.12</td>
<td>0.11</td>
</tr>
<tr>
<td><strong>RIM</strong></td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>( \kappa_{RF} )</td>
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<td>0.21</td>
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<tr>
<td>( V ) (nm(^3))</td>
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<td>78.82</td>
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<td>( r^* ) (nm)</td>
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<tr>
<td>( W(r^*) ) (kcal/mol)</td>
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<td>4.61</td>
<td>4.58</td>
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<tr>
<td>( e^{-\beta W(r^*)} )</td>
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Fig. 10 a and c: Survival probability correlation functions for \( t^* = 0.5 \text{ ps} \) at 1100 K; circles are simulation results and solid lines are bi-exponential fits. b and d: anionic escape times derived from the fits as a function of \( t^* \). a and b show that independent of the model and the choice of \( t^* \), faster chloride escape is observed for LiCl than for NaCl and KCl. This is the same trend observed in Fig. 8 for \( t \).

Obviously the choice of \( t^* \) will change \( P(t,t+\delta t,r^*) \), but sensible values should keep trends unmodified. The residence time of the \( i \)th anion is obtained from the normalized time-correlation function of the survival probability

\[
C(t,t^*) = \langle P(t,t+\delta t,r^*) \rangle_{ij} / \langle P(t,t,r^*) \rangle_{ij}.
\]

Here \( \langle \cdot \rangle_{ij} \) indicates averaging over ion pairs and time. We computed \( C(t,t^*) \) for a range of \( t^* = 100 \text{ fs} - 1 \text{ ps} \) making sure that \( t^* \) represented only short, transient escapes of anions. In other words, we chose \( t^* \) to be significantly smaller than the actual anion exchange time derived from TST and Marcus theory. Fig. 10a and Fig. 10c depict \( C(t,t^*) \) obtained for \( t^* = 500 \text{ fs} \) at 1100 K in the case of the PIM and RIM respectively. Fig. 10b and Fig. 10d show that whereas the choice of \( t^* \) affects the value of the escape times, trends are consistently the same. We analyzed our results by applying a bi-exponential fit to \( C(t,t^*) \); the fit shows a fast initial (femtoseconds) component followed by a second component in the picosecond time scale and we associate this second component with the exchange dynamics resolved by the prior two rate theory methods. The escape times presented in Fig. 10 show that anion exchange is faster for LiCl than for NaCl, which is again faster but comparable to that for KCl. While the observation is true for both simulation models, escape times are more discernible between NaNl and KCl for the RIM model. It is noticeable from Fig. 10 and Fig. 8 that except for KCl the escape times predicted by the IMM method are slightly longer than the chloride exchange timescales predicted by the Marcus and TST methods.

5 Conclusions

We have studied the liquid structure, diffusion and conductivity as well as interpreted the dynamics of anion exchange across the family of alkali chlorides in the molten state. Our X-ray scatter-
ing and simulations using different techniques emphasize that the main structural feature for these systems is charge alternation. However, when comparing $S(q)$ across these systems one finds that the lowest $q$-peak has a different origin for LiCl and KCl. In the case of LiCl the peak is truly a positive-negative charge alternation feature, whereas cancellation of peaks and antipeaks completely vanishes the charge alternation feature for KCl. A peak to the right (higher $q$) of the charge alternation peak is normally associated with the adjacent structural correlations between cations and anions, except that this peak in the case of LiCl is missing since the X-ray form factor for Li$^+$ is so small. Both the absence of a charge alternation peak in the case of KCl and the fact that the adjacency peak is missing in the case of LiCl are simply a consequence of poor contrast in the technique, and not due to a fundamental structural difference in the melt. This finding highlights the importance of computer simulations in explaining the origin of the different $S(q)$ features, without which, these assignments would have been much harder to accurately make. In general, the agreement between experiment and simulations both in real space and reciprocal space is satisfactory giving credence to predictions and analysis based on simulations. However, based on $S(q)$ and $D(r)$, it would appear that further improvements could be made to models for KCl.

The experimental and computational conductivity of these systems decreases with increasing cationic size. Two factors contribute to this decrease, the first one is the very high self contribution of Li$^+$ compared to other ions and the second is the change in molar volume across the salts. If one corrects for the change in volume by focusing instead on the molar conductivity, there is still a significant drop in going from LiCl to NaCl, but the experimental change in going from NaCl to KCl is small and computationally there is essentially no change in molar conductivity between these two. The same type of trend is observed when considering the diffusion coefficients instead of the molar conductivity.

We used three different approaches to understand the time scale and mechanisms for chloride exchange (or escape in the case of the IMM technique) across the family of cations. In all cases, we found that consistent with the trends in conductivity, the reciprocal rate of anion exchange, $\tau$, is larger for the larger cations. Different reaction coordinates and methods provide unique perspectives to the process; for example, the solvent ions couple strongly to the inter-ionic distance but less so to the coordination number. Whereas anion exchange is definitively an activated process, barriers to anion exchange do not appear to determine trends of exchange. The mass of the cation and solvent effects are salient factors in the Marcus-TST approach based on $n$ whereas the mass of the cation and the overall sample volume are important in the TST approach based on $r$. While none of the methods are superior to any other—they all generate reasonably consistent $\tau$ values bolstering confidence in our analysis—, the TST approach based on $r$ appears to provide the most intuitive link to simple factors (the cation size/mass and the sample volume) that directly correlate with trends for the conductivity of the different melts.

In the future, it may be interesting to study mixtures of these salts, particularly the low-melting eutectics which are very relevant to technology applications. We suspect that at these lower temperatures there could be a crossover between the high-temperature dynamics (where free energy barriers are not fundamentally important in defining anion exchange) and the lower-temperature regime in which free-energetics may be all that matters. We have mentioned earlier that whereas lithium has a very high mobility in molten LiCl, when mixed in a salt with a larger cation such as KCl this is reduced significantly. This phenomenon is also common when a small cation such as Li$^+$ is mixed in an room temperature ionic liquid; in this case it is understood that the hard Li$^+$ cation becomes an integral part of the ionic liquid charge network stiffening it and causing an increase in viscosity. How this happens with high temperature molten salts is something we would like to explore.

Conflicts of interest
There are no conflicts to declare.

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