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A benchmark database for mixed-solvent electrolyte solutions: Consistency analysis using E-NRTL

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8 Abstract: Modeling of thermodynamic properties of mixed-solvent electrolyte solutions is 9 challenging. Reliable experimental data are essential for any model development and parameterization. In this work, a benchmark database for (water + methanol/ethanol + alkali halide) 10 11 mixed-solvent electrolyte solutions is presented. Available experimental data of mean ionic activity 12 coefficient (MIAC) and vapor-liquid equilibrium (VLE) are comprehensively collected and critically evaluated for 61 datasets of 23 solutions. The resulting benchmark database includes 1413 13 data records from 32 datasets for 13 solutions. Evaluated datasets of the relevant aqueous 14 15 electrolyte solutions are also presented. A consistent E-NRTL model that satisfies the Gibbs-16 Duhem equation is utilized for analyzing the data, reconciling the MIAC and VLE data. The 17 collected data are critically evaluated. A benchmark database is obtained. Based on the database, 18 recommended parameters are obtained for the E-NRTL model.

Keywords: benchmark database, consistent E-NRTL, mixed-solvent electrolyte solution, mean
ionic activity coefficient, vapor-liquid equilibrium

21 **1 Introduction**

Electrolyte solutions are important in many industrial processes, e.g., CO₂ capture and sequestration ¹, flue gas treatment ², desalination ³, scale formation ⁴, corrosion resistance enhancement ⁵, batteries ⁶, pharmaceutical processes ⁷, etc. Reliable experimental data are essential for thermodynamic property model development and process system modeling, while incomplete

and inaccurate data hampers their utilization⁸. Because of the complexity of the interactions that 26 take place in electrolyte solutions, the existing thermodynamic models are not well adapted to those 27 mixtures ⁹; available tools are not yet well accepted and validated as for non-electrolyte systems ¹⁰; 28 the physics of the competing contributions is not yet well understood ^{11,12}. Furthermore, over the 29 past century, experimental data have been reported in dozens of conventions ¹³, e.g., different 30 reference states and composition units. In addition, for mixed-solvent electrolyte solutions, 31 32 experimental data from different sources are frequently not consistent with each other. Thus, obtaining a reliable database needs comprehensive collection and critical evaluation of available 33 34 experimental data.

35 In a previous work ¹⁴, a benchmark database is proposed for aqueous alkali halide solutions. For mixed-solvent electrolyte solutions, the extra solvent brings in an extra dimension in 36 composition, complexing property comparisons. Furthermore, the solvation and dissociation 37 behavior of electrolytes is very different in water and non-water solvents. Various equations of 38 state ^{15–25} and activity coefficient models ^{26–38} have been applied on these mixtures. Because of the 39 complexity from both experimental and theoretical perspectives, a benchmark database for 40 important mixtures of the type can facilitate future model development, parameterization, and 41 evaluation. In addition, compared to the data status of aqueous electrolyte solutions, that of mixed-42 solvent electrolyte solutions is much scarcer. Therefore, data analysis requires a model that applies 43 44 on mixed-solvent electrolyte systems and does not include too many adjustable parameters.

45 Although modern thermodynamic models are of theoretical basis. As parameters are regressed, accurate correlation of the experimental data does not guarantee data reliability. Therefore, critical 46 47 evaluation of the experimental data needs comparisons between data from different sources and of different properties, as well as comparisons within a series of salts in the same solvents. To achieve 48 this, a consistent version of the E-NRTL model is introduced, satisfying the Gibbs-Duhem equation, 49 and thus facilitating verification between the mean ionic activity coefficient (MIAC) and vapor-50 liquid equilibrium (VLE), which includes information of the ionic and solvent activity coefficients, 51 respectively. The consistent model is similar to the consistent model proposed by van Bochove et 52 al. ³¹, short of the Bronsted-Guggenheim term ^{39,40}. 53

54 Our aim is to present a benchmark database based on an extensive evaluation of available 55 experimental data, to recommend parameters based on this database, and to understand trends of

data and parameters. The work proceeds as follows. First, the overall data analysis framework is 56 introduced. Then, the consistent E-NRTL model is introduced and compared with the widely used 57 58 original E-NRTL. Then, a consistency analysis is performed for each mixture. Impacts of the 59 objective function and water-salt parameters on the representation of mixed-solvent electrolyte 60 solution properties are investigated, concluding on a parameter regression procedure that facilitates the data analysis framework. Based on these findings, the alcohol-salt parameters and ternary 61 62 property results are presented. Finally, the obtained benchmark database of the mixed-solvent electrolyte systems and the involved aqueous electrolyte systems is proposed. 63

64 **2 Data analysis framework**

This section introduces the properties' definitions and relationships with the activity coefficients, the composition denotations, and the raw data collected from the databanks. Figure 1 shows the data analysis framework. The following path is followed:

68 (1) We start from the raw data collected from the databanks. Datasets that are non-experimental and incomplete (e.g., datasets that are not in tabular forms, and VLE datasets without vapor 69 70 phase molar fraction) are removed. The experimental datasets were reported in many different 71 composition units. They are converted to ion-based molar fraction, i.e., full dissociation is 72 assumed, and the cation and anion are considered as two species. For the (water + alcohol) 73 mixtures, VLE data are collected. For the (water + salt) and (water + alcohol + salt) mixtures, 74 VLE and MIAC data are collected. For mixtures containing two solvents, only the VLE 75 datasets that include both vapor pressure and vapor phase molar fraction data are collected, because solvent activity coefficient requires both variables. 76

(2) Then, the binary datasets are compared against each other and with the E-NRTL model with
preliminary parameters. Doing so, inconsistent datasets are spotted as they do not agree with
data from multiple other sources (for binary mixtures, the data status is quite extensive, and
there are always datasets from multiple sources that agree with each other), and are removed.
The remaining datasets constitute the benchmark database of the binary mixtures.

82 (3) Based on these datasets, parameters are regressed for the water-alcohol and water-salt pairs.
83 Once these parameters (water-alcohol and water-salt) are sufficiently validated, they are used
84 in the evaluation of the ternary (water + alcohol + salt) datasets, along with preliminary
85 parameters for the alcohol-salt pairs. Because the solubility of salts in anhydrous alcohol is

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usually very small, in which range the model is dominated by the Pitzer-Debye-Huckel (PDH) 86 term, regression of the parameters of the alcohol-salt pairs to binary (alcohol + salt) data is 87 88 infeasible. Mixed-solvent electrolyte solution data are more suitable for obtaining alcohol-salt parameters. Therefore, the binary (alcohol + salt) mixtures are not included in this benchmark 89 90 database. Compared to the data status of the aqueous electrolyte solutions, that of the mixedsolvent electrolyte solutions is not as extensive. The datasets are evaluated by comparing with 91 92 other datasets for the same mixture when they are available, and by observing the data trends to identify obvious outliers. Inconsistent datasets are identified and removed. The remaining 93 94 datasets are further marked as "recommended", "tentative", and "uncertain", constituting the benchmark database of the mixed-solvent electrolyte solutions. Based on the database, 95 96 parameters of the E-NRTL model are regressed.



97

98 Figure 1. Data analysis framework.

99 2.1 Properties and their relationships with the activity coefficients

MIAC and VLE include information about the derivatives of the excess Gibbs energy (G^{E}) with respect to salt and solvent molar fractions, as shown in Figure 2. The arrow direction denotes that the composition derivatives of G^{E} are for the corresponding components. By definition, MIAC relates to the derivative of G^{E} over the salt composition, while VLE relates to the derivative of G^{E} over the solvent compositions. Therefore, they are ideal for model parameterization, and are included in the database.



106

107 Figure 2. Schematics of the correspondence between the properties and composition derivatives of G^{E} .

The activity coefficient describes deviation from ideality based on a given reference state and 108 composition unit. For ions, a common convention of the activity coefficient is the rational 109 110 asymmetrical activity coefficient. For aqueous electrolyte solutions, the ionic activity coefficient 111 is normalized at infinite dilution in the water solvent. For mixed-solvent electrolyte solutions, the ionic activity coefficient is normalized at infinite dilution in the solvent mixture at the same salt-112 113 free composition, because, otherwise, large experimental composition uncertainties would be present at low water composition when properties are defined in the molality unit. In the 114 experimental literature, data are always reported based on the molality unit defined in the solvent 115 mixture. 116

$$\gamma_i^* = \frac{\gamma_i}{\gamma_i^{\infty}} \tag{1}$$

117 where γ_i is the symmetrical activity coefficient, γ_i^{∞} is γ_i at infinite dilution in the solvent mixture 118 at the corresponding salt-free composition. Thus, the reference state in Eq. (1) is different according 119 to the salt-free composition of the solvent mixture. However, the rational asymmetrical activity 120 coefficient, as well as the molality and molarity conventions, are only used for presentation; while 121 γ_i itself is calculated in the models. In this work, unless otherwise noted, the reference state of ionic 122 activity coefficient is defined at infinite dilution in the solvent mixture.

In addition, the molality activity coefficient $(\gamma_i^{\rm m})$ and molarity activity coefficient $(\gamma_i^{\rm c})$ are also widely used conventions:

$$\gamma_i^{\rm m} = \gamma_i^* \sum_j^{\rm solvents} x_j \tag{2}$$

$$\gamma_i^c = \gamma_i^* \frac{\rho_{\text{solvent}}}{\rho_{\text{solution}}} \tag{3}$$

125 where ρ is the mass density. Like γ_i^* , γ_i^m and γ_i^c also take reference state at infinite dilution in the 126 solvent mixture at the corresponding salt-free composition.

127

The MIAC is usually reported in the experimental literature in the molality convention.

$$\gamma_{\pm}^{m} = (\gamma_{c}^{m\nu_{c}}\gamma_{a}^{m\nu_{a}})^{\frac{1}{\nu}}$$

$$\tag{4}$$

where $v = v_c + v_a$ is the sum of the stoichiometric coefficients, the subscript c denotes cations, and the subscript a denotes anions.

130 The MIAC is usually measured using potentiometry (also noted as electromotive force measurement)⁴¹. An alternative approach is to calculate MIAC based on the Gibbs-Duhem 131 equation, e.g., the isopiestic vapor pressure measurements conducted by Robinson and co-workers 132 for many aqueous solutions ⁴²⁻⁴⁴. However, for mixed-solvent electrolyte solutions, an infinite 133 134 dilution activity coefficient model is needed for converting the reference state of these data to the 135 infinite dilution of the salt in the solvent mixture at the same salt-free solvent composition. In this work, all the collected data are measured by potentiometry. This reflects what has been 136 137 unanimously implemented in the experimental community. In the potentiometric measurements, 138 the electromotive force of the cells are measured, and substituted into the Nernst equation as a function of molality (defined in the solvent mixture) along with an activity coefficient model, e.g., 139 the extended Debye-Hückel. With some treatment, the expression is close to linear with molality 140 ^{13,45}. Coefficients of the Nernst equation and the activity coefficient model are regressed at the same 141 time for each salt-free composition. The obtained coefficients are then used for calculating the 142 MIAC. Thus, the reported MIAC depends not only on the measurement of the particular salt 143 molality, but also on the series of measurements for the salt-free solvent composition. 144

In the experimental literature, the uncertainty of electromotive force measurement, rather than of the MIAC itself, is usually reported. The electromotive force can be measured up to very high accuracy, e.g., 0.02 mV, corresponding to 0.05% in the MIAC ⁴¹. However, the abovementioned data treatment procedure introduces uncertainty in the obtained MIAC. Furthermore, the measurement uncertainties of composition and temperature also contribute to the combined uncertainty of the MIAC.For mixed-solvent electrolyte solutions, the relatively large deviations (in some cases more than 20%) observed in mixed-solvent electrolyte measurements from different
experimental sources is a matter of reliability rather than accuracy, indicating an imperative need
for critically evaluating the data before use in model developments.

When the mixed-solvent electrolyte solution VLE data include both vapor pressure and vapor phase molar fraction, the solvent activity coefficient can be obtained taking the ideal gas assumption in the vapor phase,

$$\gamma_{i,\text{solvent}} = \frac{y_{i,\text{solvent}}p}{x_{i,\text{solvent}}p_{i,\text{solvent}}^{\text{sat}}}$$
(5)

157 where y is the molar fraction in the vapor phase, x is the molar fraction in the liquid phase, p is the pressure, $p_{i,solvent}^{sat}$ is the vapor pressure of pure solvent *i* at the temperature. Eq. (5) assumes that 158 the pressure is sufficiently small so that the vapor phase can be treated as ideal gas and that no 159 160 Poynting correction is needed in the liquid phase. Figure 3 shows the solvent activity coefficients derived from experimental VLE data for (water + ethanol + KCl) at 298.15 K using Eq. (5). As salt 161 composition increases, alcohol activity coefficient increases, while water activity coefficient 162 163 decreases. The alcohol activity coefficient changes more with the salt composition at low alcohol 164 composition, while the water activity coefficient is close to unity in the entire experimental data 165 range. As alcohol composition increases, alcohol activity coefficient decreases, while water activity 166 coefficient first decreases and then increases.



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Figure 3. Solvent activity coefficients derived from experimental VLE data ⁴⁶ for (water + ethanol + KCl)
 at 298.15 K.

The ionic and solvent activity coefficients are related to each other according to the Gibbs-Duhem equation. At constant *T* and *p*,

$$\sum n_{i,\text{solv}} d\ln x_{i,\text{solv}} \gamma_{i,\text{solv}} + n_s \sum \nu d\ln m_{\pm} \gamma_{\pm}^m = 0$$
(6)

Therefore, MIAC and VLE are complementary to each other. Experimental data of theseproperties can be utilized to verify each other.

174 **2.2** Composition denotations

Experimental data were reported in many different conventions. Typically, for mixed-solvent electrolyte mixtures, solvent compositions were usually reported in molar fraction or weight fraction on the salt-free basis, while salt compositions were usually reported in molar fraction, molality, or molarity. Before any modeling work, these data need to be converted to the same unit, typically ion-based molar fraction, which is used in most models. For instance, the composition unit of salt-free molar fraction for solvent and molality for salt can be converted to ion-based molar fraction according to,

$$x_{i,\text{solvent}} = \frac{1000 \text{ g} \times \sum_{j} \frac{x_{j,\text{solvent}}^{0}}{M_{j,\text{solvent}}}}{\nu x_{\text{salt}}^{0} + 1000 \text{ g} \times \sum_{j} \frac{x_{j,\text{solvent}}^{0}}{M_{j,\text{solvent}}}}$$

$$x_{i,\text{ion}} = \frac{\nu_{i} x_{\text{salt}}^{0}}{\nu x_{\text{salt}}^{0} + 1000 \text{ g} \times \sum_{j} \frac{x_{j,\text{solvent}}^{0}}{M_{j,\text{solvent}}}}$$
(7)
(8)

where x^0 is in the original composition unit before conversion, *M* is the molar mass, and the subscript ion denotes cation or anion. In this work, we consider only the alkali halides. Hence, v = 2, $v_i = 1$.

185 2.3 Raw data collected from databanks

Data are collected from DETHERM ⁴⁷, CERE electrolyte databank ⁴⁸, and other sources, e.g., datasets that were utilized in other modeling papers (e.g., references ^{16,19,21,23–25}), etc. All the collected data are experimental and available in public literature. Datasets that are not provided in tabular forms are excluded. VLE datasets that do not report both vapor pressure and vapor phase molar fraction are excluded. Among those mixtures, experimental data are more extensive for the mixtures containing Na⁺ salts and Cl⁻ salts. Overall, the data status of the mixed-solvent electrolyte mixtures is much less extensive compared to that of aqueous alkali halide mixtures ¹⁴. Data of the constituting (alcohol + salt) binary mixtures are not included in the database, because the data status
is even scarcer compared to that of the mixed-solvent electrolyte solutions, because that MIAC
data is almost nonexistent, and because salt solubility is very small in alcohols in most cases.

After collecting the raw database and evaluating the data quality, a final database is constructed by introducing some codes ("**R**" for "recommended", "**T**" for "tentative", and "**U**" for "uncertain"). Details will be explained in Section 6.

2.4 Statistical quantities' definitions

In what follows, the data will be compared with calculation results, resulting in the need fordefining some statistical quantities.

202 Average deviation (AD) and maximum deviation (MD): AD and MD represent deviations 203 of the model from each experimental dataset. For vapor pressure, the relative deviation is taken. For vapor phase molar fraction and MIAC, the absolute deviation is taken. The vapor phase molar 204 fraction is between 0 and 1, while MIAC is in general between 0 and 1 in the salt composition 205 206 range that is used in the parameter optimization in this work. Using the absolute deviation rather 207 than the relative deviation avoids exaggerating the deviations when the values are small. In the data analysis, x_{ion} is limited to 0.06, in which range MIAC is never much larger than 1. However, if 208 209 MIAC gets very large in the regression data, relative deviation should be used.

Objective function (OF): Parameters are optimized using a local minimizer in IFPEN's optimization software, ATOUT ⁴⁹, starting from the optimal value from 50 initial parameter sets randomly obtained in the parameter ranges using a Latin hypercube algorithm. The combined objective function (OF) is,

$$OF = \frac{1}{2} \sum_{j=1}^{n_{ds}} \frac{\sum_{i=1}^{n_{dp}^{j}} \left(\gamma_{cal}^{\pm i,j} - \gamma_{exp}^{\pm i,j} \right)^{2}}{n_{dp}^{j}} + \frac{1}{2} \sum_{j=1}^{n_{ds}} \frac{\sum_{i=1}^{n_{dp}^{j}} \left(\frac{p_{cal}^{i,j} - p_{exp}^{i,j}}{p_{exp}^{i,j}} \right)^{2}}{n_{dp}^{j}} + \frac{1}{2} \sum_{j=1}^{n_{ds}} \frac{\sum_{i=1}^{n_{ds}} \left(y_{cosolvent,cal}^{i,j} - y_{cosolvent,exp}^{i,j} \right)^{2}}{n_{dp}^{j}}$$
(9)

where n_{ds} is the number of datasets, n_{dp}^{j} is the number of data points in dataset j, the subscript cal 214 denotes calculated values, and the subscript exp denotes experimental values. Depending on the 215 216 data status, MIAC and VLE are used individually or together for the different mixtures. In the following sections, the OF that only consists of MIAC data is denoted as OF-M, the OF that only 217 218 consists of VLE data (pressure and vapor phase molar fraction) is denoted as OF-V, and the OF that consists of both MIAC and VLE is denoted as OF-MV. Quite often, uncertainty is not reported 219 in the experimental literature. In addition, the uncertainty might be reported more optimistically in 220 some experimental literature than others. Instead of using directly the data uncertainty as reported, 221 a more rigorous approach is to adjust its value based on data distribution, observed deviation after 222 correcting systematic deviations, and model capability ⁵⁰. However, the availability of experimental 223 224 data and uncertainty of the mixed-solvent electrolyte solutions does not facilitate such detailed 225 analysis. Therefore, the data uncertainty is not included in the data analysis. Instead, the same 226 weight is assigned to all the datasets that are considered to be reliable after extensive comparisons 227 between datasets from different experimental sources.

228 **3 Model for data analysis**

The electrolyte non-random two-liquid (E-NRTL) model ^{51,52} is used for data analysis because 229 of its wide acceptance in the industry ⁵³, its capacity to describe mixed-solvent electrolyte solutions 230 ²⁷, its small number of adjustable parameters, and its potential theoretical basis of parameters ⁵⁴. 231 The E-NRTL model includes a NRTL part ⁵⁵, which accounts for the local interaction contribution 232 between all interacting species, and a PDH part ⁵⁶, which accounts for the long-range Coulombic 233 interaction contribution between ions. Mock et al. ²⁶ applied the E-NRTL model on mixed-solvent 234 235 electrolyte mixtures, which was implemented in ASPEN PLUS. However, according to these 236 authors, "only the local interaction contribution term of the electrolyte NRTL model is used in this study and the long-range interaction contribution term is dropped", because their "main 237 238 consideration was the ability to represent the phase equilibrium behavior of solvent species". The model was successful in representing VLE. Liu and Watanasiri 57 introduced a Bronsted-239 Guggenheim term ^{39,40} and a Born term ^{58,59} into the model, and applied the E-NRTL model on the 240 representation of liquid-liquid equilibrium of mixed-solvent electrolyte solutions. Van Bochove et 241 242 al. ³¹ suggested that the difference in the dielectric constants of the solvents were large in the (water + organic solvent + salt) mixtures, and included the correct solvent composition derivatives in their 243

PDH and Born terms. The resulting E-NRTL model satisfies the Gibbs-Duhem equation, and thus 244 is consistent. However, they suggested that the consistent e-NRTL is only slightly better than 245 original e-NRTL. In a recent work, Chang and Lin⁶⁰ included the ion contribution to the solvent-246 related properties in their extended PDH model. However, van Bochove's derivations were largely 247 ignored in later works on mixed-solvent electrolyte modeling with the E-NRTL model. The Born 248 term ^{58,59} in the E-NRTL model accounts for the difference of medium effect between the infinite-249 dilution aqueous solution to the given mixed-solvent solution. Song and Chen²⁷ proposed a 250 segmental E-NRTL model for mixed-solvent electrolyte solutions, and later proposed a 251 symmetrical version ⁶¹ to accommodate pure fused salts. In these works, the solvent density and 252 relative permittivity were considered as pseudo-pure functions, i.e., although solvent-composition 253 254 dependence was accounted for in the function, derivative terms were omitted as activity coefficients were derived from the excess Gibbs energy. Tsanas et al. ⁶² suggested that violation of 255 256 the Gibbs-Duhem equation would result in convergence to the wrong minimum in reactive flash calculations. In this section, the consistent E-NRTL model is described; then, the consistent and 257 258 original E-NRTL models are compared to show whether the issue of consistency matters for VLE 259 and MIAC calculations, which are the scope of the presented database in this work.

260 **3.1 Consistent E-NRTL**

For mixed-solvent electrolyte solutions, the original E-NRTL ^{27,61} is usually used in a manner 261 that violates the Gibbs-Duhem equation. (The problem is not present for single-solvent systems.) 262 In Van Bochove et al.'s work ³¹, the model satisfies the Gibbs-Duhem equation, and therefore is 263 consistent. In the consistent E-NRTL model, the PDH and Born terms are different from the 264 265 original E-NRTL, as the correct solvent composition derivatives are included. This work utilizes a model that is similar to Van Bochove's consistent model, short of the Bronsted-Guggenheim term. 266 267 The model consists of a NRTL term, a PDH term, and a Born term. Here, the model is briefly introduced. 268

The activity coefficient terms are obtained by taking partial derivatives of the excess Gibbs energy terms:

$$\frac{G^{\mathrm{E,PDH}}}{RT} = -\left(\sum_{i} n_{i}\right) \frac{4A_{\varphi}I_{\chi}}{\rho} \ln\left(1 + \rho I_{\chi}^{\frac{1}{2}}\right)$$
(10)

$$\frac{G^{\text{E,Born}}}{RT} = \frac{Q_{\text{e}}^2}{2kT} \left(\frac{1}{\varepsilon_{\text{s}}\varepsilon_0} - \frac{1}{\varepsilon_{\text{w}}\varepsilon_0}\right) \sum \frac{n_i Z_i^2}{r_i}$$
(11)

where R = 8.3144598 J mol⁻¹ K⁻¹ is the universal gas constant ⁶³, *T* is the temperature, n_i is the molar number of component *i*, A_{φ} is one third of the Debye-Huckel limiting slope as given in Eq. (12), I_x is the ionic strength as given in Eq. (13), $Q_e = 1.6021766208 \times 10^{-19}$ C is the elementary charge ⁶³, $k = 1.38064852 \times 10^{23}$ m² kg s⁻² K⁻¹ is the Boltzmann constant ⁶³, ε_s is the solvent relative permittivity, ε_w is the water relative permittivity, $\varepsilon_0 = 8.854187817 \times$ 10^{-12} F m⁻¹ is the dielectric constant in the vacuum ⁶³, Z_i is the ionic charge, r_i is the Born radius and is set at 3 Å for all the ions in this work.

$$A_{\varphi} = \frac{1}{3} \left(\frac{2\pi N_{\rm A} \rho_{\rm s}}{1000} \right)^{\frac{1}{2}} \left(\frac{25 Q_{\rm e}^{\ 2}}{\pi \varepsilon_{\rm s} \varepsilon_{\rm 0} kT} \right)^{\frac{3}{2}} \tag{12}$$

$$I_x = \frac{\sum Z_i^2 n_i}{2\sum n_i}$$
(13)

278 where $N_{\rm A} = 6.022140857 \times 10^{23} \, {\rm mol}^{-1}$ is the Avogadro number ⁶³.

279 The mixing rules for solvent density and relative permittivity are,

$$\rho_{\rm s} = \frac{1}{\sum_{i}^{\rm solvents} \frac{x_i^0}{\rho_i}} \tag{14}$$

$$\varepsilon_{\rm s} = \sum_{i}^{\rm solvents} w_i^0 \varepsilon_i \tag{15}$$

280 where x_i^0 is the molar fraction on the salt-free basis, w_i^0 is the weight fraction on the salt-free basis, 281 ρ_i is the density of solvent *i*, and ε_i is the relative permittivity of solvent *i*, as given below.

$$\rho_{i} = \frac{A_{i}}{B_{i}^{1 + \left(1 - \frac{T}{C_{i}}\right)^{D_{i}}}}$$
(16)

$$\varepsilon_i = \varepsilon_i^0 + \varepsilon_i^1 \left(\frac{1}{T} - \frac{1}{273.15 \text{ K}} \right) \tag{17}$$

The coefficients, A_i , B_i , C_i , D_i , ε_i^0 and ε_i^1 , are obtained from or regressed to data from DIPPR ⁶⁴ and reference ⁶⁵, as explained in detail in the Supporting Information.

284 Thus, the activity coefficient terms are,

$$\ln \gamma_{i}^{\text{PDH}} = \left[\frac{\partial \left(\frac{G^{\text{E,PDH}}}{RT}\right)}{\partial n_{i}}\right]_{T,p,n_{j(j\neq i)}}$$

$$= \frac{G^{E,\text{PDH}}}{RT} \left\{ \frac{1}{\Sigma_{k} n_{k}} + \frac{1}{A_{\varphi}} \left(\frac{\partial A_{\varphi}}{\partial n_{i}}\right)_{T,p,n_{j(j\neq i)}}$$

$$+ \frac{1}{I_{x}} \left(\frac{\partial I_{x}}{\partial n_{i}}\right)_{T,p,n_{j(j\neq i)}} \left[1 + \frac{1}{2\left(1 + \rho^{-1}I_{x}^{-\frac{1}{2}}\right)\ln\left(1 + \rho I_{x}^{-\frac{1}{2}}\right)}\right] \right\}$$

$$\ln \gamma_{i}^{\text{Born}} = \left[\frac{\partial \left(\frac{G^{\text{E,Born}}}{RT}\right)}{\partial n_{i}}\right]_{T,p,n_{j(j\neq i)}}$$

$$= \left\{ -\frac{Q_{e}^{2}}{2kT\varepsilon_{s}^{2}\varepsilon_{0}} \left(\frac{\partial \varepsilon_{s}}{\partial n_{i}}\right)_{T,p,n_{j(j\neq i)}} \left(\sum \frac{n_{k}Z_{k}^{2}}{100r_{k}}\right) \text{ for solvents} \right\}$$

$$\left. \frac{Q_{e}^{2}}{2kT} \left(\frac{1}{\varepsilon_{s}\varepsilon_{0}} - \frac{1}{\varepsilon_{w}\varepsilon_{0}}\right) \frac{Z_{i}^{2}}{100r_{i}} \text{ for ions} \right\}$$

$$(18)$$

where the analytical expressions for $\left(\frac{\partial A_{\varphi}}{\partial n_i}\right)_{T,p,n_{j(j\neq i)}}$ and the other derivatives over n_i are provided

in the Supporting Information.

The 2nd term in Eq. (18) and the upper equation in Eq. (19) are not included in the original E-NRTL, as derivatives of solvent density and relative permittivity over solvent composition are ignored. As these terms are included, the Gibbs-Duhem equation, which is violated by the original E-NRTL, is satisfied. Consequently, in principle, one can expect that the consistent E-NRTL can reconcile the ion and solvent properties (MIAC and VLE) better.

The consistent E-NRTL model uses exactly the same formulation for the NRTL term as the original E-NRTL ^{51,52}.

$$\ln \gamma_{\rm m}^{\rm NRTL} = \frac{\sum_{j} x_{j} G_{j\rm m} \tau_{j\rm m}}{\sum_{j} x_{j} G_{j\rm m}} + \sum_{\rm m'} \frac{x_{\rm m'} G_{\rm mm'}}{\sum_{k} x_{k} G_{k\rm m'}} \left(\tau_{\rm mm'} - \frac{\sum_{k} x_{k} G_{k\rm m'} \tau_{k\rm m'}}{\sum_{k} x_{k} G_{k\rm m'}} \right) + \sum_{\rm c} \frac{\sum_{a'} \frac{x_{a'} x_{\rm c} G_{\rm mc,a'c}}{\sum_{k} x_{k} G_{k\rm c,a'c}} \left(\tau_{\rm mc,a'c} - \frac{\sum_{k \neq \rm c} x_{k} G_{k\rm c,a'c} \tau_{k\rm c,a'c}}{\sum_{k \neq \rm c} x_{k} G_{k\rm c,a'c}} \right)}{\sum_{a''} x_{a''}} + \sum_{\rm a} \frac{\sum_{\rm c'} \frac{x_{\rm c'} x_{a} G_{\rm ma,c'a}}{\sum_{k} x_{k} G_{k\rm a,c'a}} \left(\tau_{\rm ma,c'a} - \frac{\sum_{k \neq \rm a} x_{k} G_{k\rm a,c'a} \tau_{k\rm a,c'a}}{\sum_{k \neq \rm a} x_{k} G_{k\rm a,c'a}} \right)}{\sum_{c''} x_{c''}}$$
(20)

$$\frac{\ln \gamma_{\rm c}^{\rm NRTL}}{Z_{\rm c}} = \frac{\sum_{\rm a'} \frac{x_{\rm a'} \sum_{\rm k} x_{\rm k} G_{\rm kc,a'c} \tau_{\rm kc,a'c}}{\sum_{\rm k} x_{\rm k} G_{\rm kc,a'c}} + \sum_{\rm m} \frac{x_{\rm m} G_{\rm cm}}{\sum_{\rm k} x_{\rm k} G_{\rm km}} \left(\tau_{\rm cm} - \frac{\sum_{\rm k} x_{\rm k} G_{\rm km} \tau_{\rm km}}{\sum_{\rm k} x_{\rm k} G_{\rm km}}\right) + \sum_{\rm m} \frac{\sum_{\rm c'} x_{\rm m} G_{\rm cm}}{\sum_{\rm k} x_{\rm k} G_{\rm km}} \left(\tau_{\rm cm} - \frac{\sum_{\rm k} x_{\rm k} G_{\rm km} \tau_{\rm km}}{\sum_{\rm k} x_{\rm k} G_{\rm km}}\right) + \sum_{\rm m} \frac{\sum_{\rm c'} \frac{x_{\rm c'} x_{\rm a} G_{\rm ca,c'a}}{\sum_{\rm k} x_{\rm k} G_{\rm ka,c'a}} \left(\tau_{\rm ca,c'a} - \frac{\sum_{\rm k \neq \rm a} x_{\rm k} G_{\rm ka,c'a} \tau_{\rm ka,c'a}}{\sum_{\rm k \neq \rm a} x_{\rm k} G_{\rm ka,c'a}}\right)}$$
(21)

$$\frac{\ln \gamma_{a}^{\text{NRTL}}}{Z_{a}} = \frac{\sum_{c'} \frac{x_{c'} \sum_{k} x_{k} G_{ka,c'a} \tau_{ka,c'a}}{\sum_{k} x_{k} G_{ka,c'a}}}{\sum_{c''} x_{c''}} + \sum_{m} \frac{x_{m} G_{am}}{\sum_{k} x_{k} G_{km}} \left(\tau_{am} - \frac{\sum_{k} x_{k} G_{km} \tau_{km}}{\sum_{k} x_{k} G_{km}} \right) + \sum_{c} \frac{\sum_{a'} \frac{x_{a'} x_{c} G_{ac,a'c}}{\sum_{k} x_{k} G_{kc,a'c}} \left(\tau_{ac,a'c} - \frac{\sum_{k \neq c} x_{k} G_{kc,a'c} \tau_{kc,a'c}}{\sum_{k \neq c} x_{k} G_{kc,a'c}} \right)}{\sum_{a''} x_{a''}}$$
(22)

 $\tau_{\rm ma,ca} = \tau_{\rm am} - \tau_{\rm ca,m} + \tau_{\rm m,ca} \tag{23}$

$$\tau_{\rm am} = -\ln \frac{G_{\rm am}}{\alpha_{\rm am}} \tag{24}$$

$$G_{\rm am} = \frac{\sum_{\rm c} x_{\rm c} G_{\rm ca,m}}{\sum_{\rm c} x_{\rm c}}$$
(25)

$$\alpha_{\rm am} = \frac{\sum_{\rm c} x_{\rm c} \alpha_{\rm ca,m}}{\sum_{\rm c} x_{\rm c}} \tag{26}$$

$$\tau_{\rm mc,ac} = \tau_{\rm cm} - \tau_{\rm ca,m} + \tau_{\rm m,ca} \tag{27}$$

$$\tau_{\rm cm} = -\ln \frac{G_{\rm cm}}{\alpha_{\rm cm}} \tag{28}$$

$$G_{\rm cm} = \frac{\sum_{\rm a} x_{\rm a} G_{\rm ca,m}}{\sum_{\rm a} x_{\rm a}} \tag{29}$$

$$\alpha_{\rm cm} = \frac{\sum_{\rm a} x_{\rm a} \alpha_{\rm ca,m}}{\sum_{\rm a} x_{\rm a}} \tag{30}$$

$$G = e^{-\alpha\tau} \tag{31}$$

Eqs. (18), (19), (21), and (22) are in the symmetrical convention. For ions, they can be converted into the rational unsymmetrical and molality conventions according to Eqs. (1) and (2).

297 The α parameters are interchangeable, i.e., $\alpha_{ii} = \alpha_{ii}$. In this work, there are only one cation and one anion in each mixture. Thus, the salt-salt parameters, $\tau_{ca,c'a} = \tau_{ca,c'}$ and $\tau_{ac,a'c} = \tau_{ac,a'}$, are 298 not relevant here. The adjustable parameters are $\alpha_{water,alcohol}$, $\alpha_{water,salt}$, $\alpha_{alcohol,salt}$, $\tau_{water,alcohol}$, 299 300 $\tau_{alcohol,water}$, $\tau_{water,salt}$, $\tau_{salt,water}$, $\tau_{alcohol,salt}$, and $\tau_{alcohol,salt}$. In short, these parameters are noted as α_{wa} , α_{ws} , α_{as} , τ_{wa} , τ_{aw} , τ_{ws} , τ_{sw} , τ_{as} , and τ_{as} . $\alpha_{wa} = 0.3$, $\alpha_{ws} = 0.2$, and $\alpha_{as} = 0.2$ are taken according to common 301 practice ²⁷. There is no adjustable parameter in the PDH and Born terms. Therefore, the remaining 302 adjustable parameters are the τ parameters. In addition, a temperature dependence is introduced for 303 304 τ:

$$\tau = \tau_0 + \tau_1 \left(\frac{1}{T} - \frac{1}{T_0} \right)$$
(32)

where τ_0 is obtained at T_0 , at which isothermal datasets are usually available, typically at 298.15 K, while τ_1 is obtained with datasets at other temperatures.

307 **3.2 Is consistency an issue?**

Because that the difference between the consistent and original E-NRTL models is only with 308 309 solvent composition derivatives, and that MIAC only involves ion composition derivatives, the consistent modification has no impact on MIAC. Thus, the parameters of the consistent and original 310 311 E-NRTL models are identical for OF-M. Table 1 shows the ADs and MDs of the consistent and original E-NRTL models with OF-M and OF-MV. For vapor pressure, the relative deviation is 312 taken (see Eq. (9)). For vapor phase molar fraction and MIAC, the absolute deviation is taken, 313 314 because these data are in general between 0 and 1, and thus the relative deviations are exaggerated 315 when the values are small. In the cases with OF-M, VLE is predicted. For the methanol mixtures, with OF-M, the consistent E-NRTL model predicts both VLE p and y more accurately than the 316 317 original model; with OF-MV, there is no significant difference between the consistent and original 318 E-NRTL models. For the ethanol mixtures, however, with either OF-M or OF-MV, the difference

- 319 between the consistent and original E-NRTL models is not significant in terms of the ADs and
- 320 MDs of MIAC and VLE.

Table 1. Average deviations (ADs) and maximum deviations (MDs) of the consistent and original E-NRTL models with OF-M (MIAC) and OF-MV (MIAC and

322 VLE). The prediction deviations are presented in italic and blue. Temperature and composition ranges of the experimental datasets are provided in Table 4. Ion

323 composition is cut off at $x_{ion} = 0.06$.

Mixture	Property	Reference	Consistent, OF-M		Consisten	Consistent, OF-MV Original, OF-M		Original, OF-MV		
			100AD	100MD	100AD	100MD	100AD	100MD	100AD	100MD
(water + methanol + NaCl)	MIAC	Xu et al. 2014 66	0.56	1.4	0.60	1.4	0.56	1.4	0.64	1.5
		Basili et al. 1996 ⁶⁷	1.4	5.0	1.5	6.0	1.4	5.0	1.5	6.7
	VLE	Yang & Lee 1998 68, p	6.0	12	2.2	5.5	8.2	14	2.2	5.5
		у	2.8	6.2	1.1	2.3	4.2	8.0	1.1	2.3
(water + methanol + KCl)	MIAC	Basili et al. 1997 ⁶⁹	1.1	4.2	1.2	4.5	1.1	4.2	1.2	4.6
	VLE	Yang & Lee 1998 68, p	3.3	7.4	1.8	4.3	4.5	8.7	1.7	4.3
		у	1.4	3.1	0.48	0.91	2.1	4.1	0.46	0.86
(water + ethanol + NaCl)	MIAC	Esteso et al. 1989 ⁷⁰	2.4	7.2	2.4	7.2	2.4	7.2	2.5	7.3
		Mamontov et al. 2016 ^{71 a}	0.91	2.7	0.91	2.9	0.91	2.7	1.0	3.5
	VLE	Yang et al. 1979 ⁴⁶ , p	2.1	5.5	2.2	6.5	1.8	4.4	2.0	6.0
		у	1.9	3.4	1.6	3.3	2.8	4.3	1.8	3.5
(water + ethanol + KCl)	MIAC	Mussini et al. 1995 ⁷²	0.88	1.7	0.88	1.8	0.88	1.7	0.90	1.7
	VLE	Yang et al. 1979 46 , p	2.0	3.7	2.4	4.0	0.98	2.6	2.3	4.0
		у	1.9	2.8	1.7	2.6	2.6	3.7	1.8	2.8

324 Note: ^a The dataset includes data at other temperatures. Deviations in this table are for the 298.15 K part of the dataset.

325 However, the more significant differences in trends are masked by the AD and MD values. Figure 4 shows the comparison of the solvent activity coefficients of (water + methanol + NaCl) 326 327 calculated using the consistent and original E-NRTL models with OF-M and OF-MV. The line types denote the different models. The colors denote the salt compositions. The lines and dots in 328 the upper part of the graph are alcohol activity coefficient, while those in the lower part are water 329 activity coefficient. The latter is close to unity for all the mixtures in the entire data ranges, and 330 331 does not represent a good criterion against the models. Thus, here the analysis focuses on the alcohol activity coefficient. Figures for (water + methanol + KCl), (water + ethanol + NaCl/KCl) 332 333 are provided in the Supporting Information. With OF-MV, the results calculated with the consistent 334 and original E-NRTL models almost overlap. With OF-M, the results calculated with the consistent 335 E-NRTL model is closer to the experimental data compared to the original model for all the mixtures. Therefore, although not to the level of the OF-MV results, VLE is predicted more 336 337 accurately with the consistent model compared to the original model, when only MIAC is available, 338 which is often the case as shown in the data summary in Section 6. A table is provided in the 339 Supporting Information for the parameters. For all the mixtures, the parameters regressed with OF-340 M and OF-MV are closer for the consistent model compared to the original, which confirms that the MIAC and VLE properties are better reconciled when calculated using the consistent E-NRTL 341 model. In addition, the alcohol activity coefficient of the methanol mixtures present larger 342 343 deviations with OF-M. Figure 5 shows the vapor pressure of (water + methanol + NaCl) with 10% and 30% weight fraction methanol (salt-free basis) with OF-M and OF-MV at 298.15 K. As salt 344 composition increases, the experimental vapor pressure increases, i.e., the salting out behavior 345 346 dominates vapor pressure change even at this low alcohol composition; the consistent and original E-NRTL models with OF-MV capture this behavior; while the models with OF-M fail to capture 347 348 this behavior, which results in the comparatively larger deviations as shown in Table 1.



349

Figure 4. Comparison of the solvent activity coefficients of (water + methanol + NaCl) calculated using

- the consistent and original E-NRTL models with OF-M and OF-MV at 298.15 K. The upper part of the
- 352 graph is alcohol activity coefficient, while the lower part is water activity coefficient. The experimental
- $353 \qquad \text{data is from reference}^{68}.$



354

Figure 5. Vapor pressure of (water + methanol + NaCl) with 10% and 30% weight fraction methanol (saltfree basis) with OF-M and OF-MV at 298.15 K. The experimental data is from reference 68 .

To sum up, with OF-MV, there is no significant difference between the consistent and original E-NRTL models; with OF-M, the consistent E-NRTL model is more accurate than the original model, in terms of both VLE results and trends of solvent activity coefficients. In the following context, unless specifically noted, the consistent E-NRTL model is analyzed; "E-NRTL" refers to the consistent model.

362 3.3 E-NRTL at larger salt composition

This section investigates how the E-NRTL model extrapolates to larger salt composition. The aim is to find a cut-off salt composition that can be used in data selection, rather than to obtain accurate results in as large range as possible. We find that the E-NRTL model is not accurate for
both aqueous and mixed-solvent electrolyte solutions at large salt composition.

367 Results of (water + salt) binary mixtures are shown rather than those of (water + alcohol + salt) ternary mixtures, because the former are confirmed by experimental data from various sources. 368 Figures 6 and 7 show the deviations of vapor pressure calculated with the E-NRTL model with OF-369 MV and parameters regressed in the region up to $x_{ion} = 0.06$ and in the entire x_{ion} range for (water 370 + NaCl) and (water + LiCl). Results for MIAC is not significantly different within $x_{ion} = 0.06$ and 371 is presented in the Supporting Information. References for the experimental datasets are provided 372 373 in the Supporting Information. The red and blue dots are the same datasets compared with the E-NRTL model using different parameters. In this work, all electrolyte solution data (MIAC and VLE, 374 375 aqueous and mixed-solvent) are cut off at $x_{ion} = 0.06$ when parameters are regressed. There are 376 three scenarios following this cut off. For salts with low solubility, e.g., KCl and NaF, the cut off covers all or most of the solubility range. Thus, the cut off x_{ion} is close to the solubility limit, i.e., 377 the parameters regressed up to $x_{ion} = 0.06$ applies up to the solubility limit. For salts with 378 solubility that is moderately larger than $x_{ion} = 0.06$, e.g., NaCl, MIAC and VLE calculated with 379 parameters regressed up to $x_{ion} = 0.06$ are slightly more accurate at lower salt composition 380 381 compared to those calculated with parameters regressed in the entire x_{ion} range (approximately up to $x_{ion} = 0.09$); however, deviations increase drastically beyond $x_{ion} = 0.06$ and reach 0.15 at 382 $x_{\text{ion}} = 0.09$, while the deviation of MIAC calculated with parameters regressed in the entire x_{ion} 383 range is within 0.05 at $x_{ion} = 0.09$; for vapor pressure, the increase of deviation is much smaller 384 in the large x_{ion} range compared to that of MIAC, being slightly larger than 3% at the solubility 385 limit. When regressed to MIAC and VLE data in the entire composition range, the obtained 386 parameters are exactly the same as those reported by Yan and Chen ⁷³, i.e., $\tau_{sw} = -4.54$, $\tau_{ws} = 8.86$. 387 For salts with solubility that is much larger than $x_{ion} = 0.06$, e.g., LiCl, however, although VLE 388 is included in the objective function along with MIAC, regression in the entire x_{ion} range results 389 390 in large deviation in vapor pressure (up to 50%) in both low and high salt composition ranges. To sum up, regression to the entire x_{ion} range can improve accuracy in the high salt composition range 391 for salts with solubility that is moderately larger than $x_{ion} = 0.06$, but cannot for salts with 392 solubility that is much larger than $x_{ion} = 0.06$ (e.g., LiCl); in both cases, accuracy in the low salt 393 394 composition range is worse compared to that calculated using parameters regressed up to $x_{ion} =$

395 0.06. It is necessary to account for model capability when selecting the range of data used in 396 parameter regression rather than use everything available. Furthermore, salts present very different 397 solubility, while cutting off at certain x_{ion} provides a common ground that facilitates finding 398 parameter trends. Therefore, although the deviations are larger at the high salt composition for the 399 (water + NaCl) case, it is still decided that all data are cut off at $x_{ion} = 0.06$. The model is here 400 used to obtain a guide for data selection, rather than to obtain accurate results in large ranges.



401

402 Figure 6. Deviations from experimental data of vapor pressure calculated with the E-NRTL model with

403 OF-MV and parameters regressed in the region up to $x_{ion} = 0.06$ and in the entire x_{ion} range for (water +

404 NaCl). References for the experimental datasets are provided in the Supporting Information.



405

406 Figure 7. Deviations from experimental data of vapor pressure calculated with the E-NRTL model with

- 407 OF-MV and parameters regressed in the region up to $x_{ion} = 0.06$ and in the entire x_{ion} range for (water +
- 408 LiCl). References for the experimental datasets are provided in the Supporting Information.

409 **4 Data consistency analysis**

410 This section presents results for the data consistency analysis. The impact of the objective

411 functions (Section 4.1) and the selection of water-salt parameters (Section 4.2) are discussed. Then,

the determination of the alcohol-salt parameters and the ternary results (Section 4.3) are shown.

413 **4.1 Impact of the objective functions**

414 In this section, the impact of the objective functions are analyzed. OF-M, OF-V, and OF-MV are compared. In Table 2, the three objective functions are tested on (water + ethanol + NaCl). The 415 416 resulting deviations on MIAC and VLE are evaluated. For OF-M, VLE results are predicted. For OF-V, MIAC results are predicted. The analysis is performed with two sets of water-salt parameters 417 418 and different upper and lower bounds for the alcohol-salt parameters. Let us first focus on the 1st water-salt parameter set, which is the recommended set. The 2nd parameter set will be discussed 419 420 in Section 4.2. Alcohol-salt parameters are optimized in two ranges [(-5, -1), (4, 12)] and [(4, 16), (4, 12)](4, 8)] (extended when boundaries are reached in optimizations), which are found to be reasonable 421 422 ranges using a trial-and-error procedure.

423 One can observe that different optimal parameter sets could be found when different parameter ranges are imposed, indicating possible local minima in the OFs. In Table 2, results 424 425 calculated with OF-M and OF-MV are approximately as good, only slightly better with the 426 parameter set obtained in [(-5, -1), (4, 12)]. However, OF-V results in large deviations from the 427 MIAC datasets, when parameters are regressed in both ranges. Therefore, in the cases that VLE is 428 not available, VLE predictions are acceptable when using parameters regressed only based on 429 MIAC; however, in the cases that MIAC is not available, MIAC predictions are likely to be far off 430 when using parameters regressed only based on VLE.

Table 2. Alcohol-salt parameters regressed in different ranges and deviations of the E-NRTL model for (water + ethanol + NaCl) with two sets of water-salt parameters (1st set: $\tau_{sw} = -4.2915$, $\tau_{ws} = 8.2464$; 2nd set: $\tau_{sw} = -1.0391$, $\tau_{ws} = -4.4583$). The information provided in the column on the left of the parameter values are the ranges in which the parameters are regressed. AD stands for average deviation. MD stands for maximum deviation. Predicted values are shown in italic and blue. Objective functions are as defined in Section 2.4.

$ au_{ m wa}$ & $ au_{ m aw}$	OF	$ au_{ m sa}$		$ au_{ m as}$		α	Esteso e	t al. ⁷⁰	Mamont	ov et al. 71	Yang et	al. ⁴⁶ p	у	
							100AD	100MD	100AD	100MD	100AD	100MD	100AD	100MD
1st	М	(-5, -1)	- 1.8983	(4, 12)	7.6887	0.2	2.4	7.2	0.91	2.7	2.1	5.5	1.9	3.4
		(4, 16)	15.978	(4, 8)	6.0436	0.2	2.5	7.4	1.1	3.7	4.7	12	1.4	3.5
	V	(-5,1)	- 0.12873	(4, 12)	5.3666	0.2	3.9	8.8	3.9	8.8	2.1	6.6	1.6	3.3
		(4, 16)	15.996	(4, 8)	4.3070	0.2	5.9	19	7.5	21	2.1	6.3	1.6	3.3
	MV	(-5, -1)	- 1.4847	(4, 12)	7.2184	0.2	2.4	7.2	0.91	2.9	2.2	6.5	1.6	3.3
		(4, 16)	16.000	(4, 8)	5.9058	0.2	2.5	7.4	1.2	3.0	4.5	11	1.4	3.3
2nd	М	(1, 5)	3.4910	(-5,1)	- 3.1604	0.2	3.3	8.9	1.6	4.2	1.3	3.3	2.2	3.7
	V	(1, 5)	3.2014	(-5,1)	- 1.3305	0.2	3.3	8.9	1.7	4.3	2.0	5.5	1.8	3.5
	MV	(1, 5)	3.5471	(-5,1)	- 1.8342	0.2	3.3	8.9	1.7	4.2	1.9	5.1	1.9	3.5

435

436 **4.2 Selection of water-salt parameters**

437 Modeling the aqueous electrolyte solutions is "simple". Usually, a few sets of parameters result in approximately the same accuracy level. However, these parameters are not as good when 438 439 extended to mixed-solvent electrolyte solutions. In Table 2, an example is provided for the (water 440 + ethanol + NaCl) case. The optimal alcohol-salt parameters and ternary results calculated with the 441 E-NRTL model and two sets of water-salt parameters are presented. The 1st water-salt parameter set is ($\tau_{sw} = -4.2915$, $\tau_{ws} = 8.2464$). The 2nd water-salt parameter set is ($\tau_{sw} = -1.0391$, 442 $\tau_{\rm ws} = -4.4583$). The 1st water-salt parameter set is regressed in a range that is close to the values 443 reported by Chen's group ^{27,54}. The 2nd parameter set is regressed in a different range. They are 444 445 approximately as accurate for representing VLE and MIAC of the (water + NaCl) binary mixture. 446 However, for the (water + ethanol + NaCl) ternary mixture, for both OF-M and OF-MV, the 2nd parameter set is less accurate for representing ternary MIAC, although approximately as accurate 447 for the VLE dataset. Furthermore, the alcohol-salt parameter set is very different for the two water-448 449 salt parameter sets. Thus, the optimal parameters for the water-salt and alcohol-salt pairs are not 450 independent. Therefore, in this work, parameters are regressed near the water-NaCl parameters as a first attempt for all the other water-salt pairs, which turns out to work very well. 451

452 **4.3 Alcohol-salt parameters and ternary results**

453 Table 3 shows the regressed parameters of the E-NRTL model for the (water + 454 methanol/ethanol + alkali halide) mixtures. Parameters are regressed according to Eq. (9). Wateralcohol parameters are regressed to OF-V. Water-salt and alcohol-salt parameters are regressed to 455 456 OF-M or OF-MV, depending on the data availability and reliability. The used (water + salt) datasets are summarized in the Supporting Information. Information of the used (water + alcohol + salt) 457 datasets are summarized in Table 4. In many cases, reliable MIAC data are available at 298.15 K. 458 Thus, $\tau_{0,solvent-salt}$ and $\tau_{0,salt-solvent}$ are first regressed; then, if reliable data are available at other 459 temperatures, the temperature gradients, $\tau_{1,solvent-salt}$ and $\tau_{1,salt-solvent}$, are regressed. When data 460 are not available at 298.15 K, the τ_0 and τ_1 parameters are regressed together. The water-salt 461 462 parameters are regressed to (water + salt) data, while the alcohol-salt parameters are regressed to 463 (water + alcohol + salt) data. As we have shown in Section 4.1, OF-M and OF-MV can be used in 464 parameter regression, while OF-V results in large deviations in MIAC predictions. For most of the 465 listed mixtures, MIAC data are available. Solvent-salt parameters that are regressed to OF-M and 466 OF-MV are marked as bold in the table. In some cases, only VLE data are available; the solvent-467 salt parameters are regressed to OF-V, and are marked as green in the table. In a few cases for the mixed-solvent electrolyte mixtures, all the available datasets are decided to be unreliable after 468 evaluation; the parameters are marked as blue and italic in the table. In some cases, data are only 469 available at 298.15 K; no temperature gradient is given in the table. No trend is observed within 470 471 the parameters. However, for all the water-salt pairs, the parameters are within a small range; for 472 the alcohol-salt pairs, the parameters are more scattered, likely because of the lower consistency of the experimental datasets of the (water + alcohol + salt) ternary mixtures compared to those of the 473 474 (water + salt) binary mixtures.

475 The overall average deviations (ADs) of the E-NRTL model from experimental datasets are presented in Table 5 along with results calculated using the other parameterization strategies. The 476 477 ADs and maximum deviations (MDs) for each dataset are presented in the Supporting Information. Deviations are only shown for the mixtures for which the parameters that are regressed to OF-M 478 479 and OF-MV. Temperature and composition ranges of the entire dataset are also shown. The datasets are only compared up to $x_{ion} = 0.06$ and T = 400 K. The behavior of the model beyond the 480 composition range has been discussed in Section 3.3. The impact of temperature is captured by the 481 482 temperature gradient introduced in Eq. (32), by the temperature dependence of the pure fluid 483 density and relative permittivity, and by the temperature dependence in the model formulation. The 484 two temperature gradient parameters do not present any trend, even though VLE data at different 485 temperatures are included in the regressions. Parameter degeneracy is observed between the two 486 temperature gradients. Figure 8 shows the comparison of MIAC calculated using the E-NRTL 487 model and the experimental data for (water + ethanol + NaCl). The agreement between the 488 experimental datasets from different sources, and between the datasets and the E-NRTL model, 489 confirms the reliability of the evaluated experimental data, as well as that of the regressed 490 parameters.

491 Table 3. Parameters of the E-NRTL model for the (water + methanol/ethanol + alkali halide) mixtures.

492 Water-alcohol parameters are regressed to OF-V. Water-salt and alcohol-salt parameters are regressed to

493 OF-M or OF-MV, depending on the data availability and reliability. The used (water + salt) datasets are

494 summarized in the Supporting Information. The relevant (water + alcohol + salt) datasets are summarized in Table 4. Solvent-salt parameters that are regressed to OF-M and OF-MV are marked as bold. 495

496

Parameters that are determined based on unreliable (uncertain) datasets are marked as italic and blue. 497 Solvent-salt parameters that are regressed to OF-V are marked as green.

Binary pair (<i>i-j</i>)	α	A			
		$ au_{0,ji}$	$ au_{0,ij}$	$ au_{1,ji}$	$ au_{1,ij}$
Water-methanol	0.3	0.21334	0.29399	1175.2	- 1900.8
Water-ethanol	0.3	0.079833	1.4142	347.97	- 634.43
Solvent-salt paramete	ers that are	regressed to OI	F-M and OF-M	V.	
Water-LiCl	0.2	- 4.9031	9.3612	- 299.61	608.71
Water-NaCl	0.2	- 4.2915	8.2464	- 111.94	604.96
Water-KCl	0.2	- 4.0036	7.8180	- 550.87	1628.3
Water-RbCl	0.2	- 4.0639	8.0359	- 20.506	334.38
Water-CsCl	0.2	- 4.3330	8.7956	- 102.70	424.63
Water-NaF	0.2	- 4.5064	9.0168	/	1
Water-LiBr	0.2	- 5.0894	9.7570	685.09	- 2848.1
Water-NaBr	0.2	- 4.5219	8.6862	20.843	122.11
Water-KBr	0.2	- 4.0552	7.8781	332.36	- 998.97
Water-CsBr	0.2	- 4.2563	8.6560	53.154	- 212.31
Water-NaI	0.2	- 4.6495	8.8420	- 101.08	434.72
Water-KI	0.2	- 3.9659	7.5374	- 83.600	250.09
Methanol-LiCl	0.2	- 4.2412	9.0090	- 755.47	4409.9
Methanol-NaCl	0.2	- 2.4187	6.8712	563.09	1775.6
Methanol-KCl	0.2	- 2.6033	7.7412	- 1350.1	5212.9
Methanol-RbCl	0.2	- 2.8102	7.5196	/	1
Methanol-NaF	0.2	- 4.7905	15.000	/	1
Ethanol-NaCl	0.2	- 1.4847	7.2184	- 19.615	2077.0
Ethanol-KCl	0.2	- 0.10247	6.9013	2211.8	3698.4
Ethanol-CsCl	0.2	- 1.5844	7.4986	/	1
Ethanol-NaF	0.2	1.5302	14.980	1	1
Parameters that are de	etermined	based on unrelia	able (uncertain)) datasets.	
Methanol-CsCl	0.2	-3.3776	9.5764	/	/

Binary pair (<i>i-j</i>)	α	А			
		$ au_{0,ji}$	$ au_{0,ij}$	$ au_{1,ji}$	$ au_{1,ij}$
Methanol-NaBr	0.2	- 3.2938	6.8686	/	/
Methanol-CsBr	0.2	-4.8626	13.080	/	/
Ethanol-LiCl	0.2	-2.9328	7.9705	/	/
Ethanol-NaBr	0.2	-1.6454	6.4317	2788.69	- 374.35
Solvent-salt parameter	ers that are	regressed to Ol	F-V.		
Ethanol-KBr	0.2	- 2.9567	6.0368	- 646.38	642.01
Ethanol-NaI	0.2	- 3.4167	7.0721	/	/
Ethanol-KI	0.2	- 1.5412	1.9649	- 703.52	- 952.12

Table 4. Summary of the experimental datasets for the (water + methanol/ethanol + alkali halide) mixtures. The significance of the letters under "group" is explained in Section 5.2.

Reference	Group	<i>T</i> (K)	$Max(x_{ion})$	w^0_{alc}	$N_{\rm DP}$
(water + methanol + LiCl)					
MIAC					
Harned 1962 ⁷⁴	Т	298.15	0.037	0.100 - 0.200	18
Mussini et al. 2000 ⁴⁵	Т	298.15	0.131	0.200 - 0.800	128
Basili et al. 1999 ⁷⁵	Т	298.15	0.131	0.200	12
Hu et al. 2008 ⁷⁶	Т	298.15	0.027	0.050 - 0.150	46
VLE					
Broul et al. 1969 77	Т	333.15	0.179	0.010 - 0.892	40
(water + methanol + NaCl)					
MIAC					
Xu et al. 2014 66	R	298.15	0.022	0.100	28
Basili et al. 1996 ⁶⁷	R	298.15	0.068	0.200 - 0.800	76
Yao et al. 1999 78	R	308.15 - 318.15	0.038	0.100 - 0.900	206
Hernandez-Hernandez et al. 2007 ⁷⁹	R	308.15	0.035	0.100	17
VLE					
Yang & Lee 1998 68	R	298.15	0.065	0.100 - 0.300	7
Yao et al. 1999 80	R	318.15	0.090	0.110 - 0.949	30
Johnson & Furter 1960 ⁸¹	R	339.25 - 352.15	0.092	0.050 - 0.928	12
(water + methanol + KCl)					
MIAC					
Basili et al. 1997 ⁶⁹	R	298.15	0.037	0.200 - 0.600	47

Reference	Group	<i>T</i> (K)	$Max(x_{ion})$	w^0_{alc}	$N_{\rm DP}$
Harned 1962 ⁷⁴	Т	298.15	0.037	0.100 - 0.200	18
VLE					
Yang & Lee 1998 68	R	298.15	0.029	0.100 - 0.300	6
Johnson & Furter 1960 ⁸¹	R	338.45 - 351.55	0.095	0.042 - 0.944	12
(water + methanol + RbCl)					
MIAC					
Zhang et al. 2004 ⁸²	R	298.15	0.043	0.100	9
Basili et al. 1997 ⁶⁹	R	298.15	0.068	0.200 - 0.800	67
(water + methanol + CsCl)					
MIAC					
Hu et al. 2007 ⁸³	U	298.15	0.029	0.100 - 0.400	72
Cui et al. 2007 ⁸⁴	U	298.15	0.061	0.100 - 0.400	64
Falciola et al. 2006 ⁸⁵	U	298.15	0.018	0.250 - 0.750	54
(water + methanol + NaF)					
MIAC					
Hernandez-Luis et al. 2003 ⁸⁶	R	298.15	0.010	0.100 - 0.900	113
VLE					
Boone et al. 1976 ⁸⁷	U	338.95 - 368.65	0.028	0.047 - 0.953	25
(water + methanol + NaBr)					
MIAC					
Han & Pan 1993 ⁸⁸	U	298.15	0.064	0.100 - 0.900	108
VLE					
Yang & Lee 1998 68	Т	298.15	0.097	0.236 - 0.423	5
Boone et al. 1976 ⁸⁷	U	338.95 - 370.55	0.095	0.042 - 0.959	23
(water + methanol + CsBr)					
MIAC					
Falciola et al. 2006 ⁸⁵	U	298.15	0.018	0.250 - 0.750	54
(water + methanol + CsI)					
MIAC					
Falciola et al. 2006 ⁸⁵	U	298.15	0.018	0.250 - 0.750	54
(water + ethanol + LiCl)					
MIAC					
Hu et al. 2008 ⁷⁶	U	298.15	0.026	0.050 - 0.150	42

Reference	Group	<i>T</i> (K)	$Max(x_{ion})$	w^0_{alc}	$N_{\rm DP}$
Hernandez-Luis et al. 2008 ⁸⁹	U	298.15	0.097	0.200 - 0.800	64
VLE					
Shaw & Butler 1930 90	U	298.15	0.133	0.148 - 0.992	24
(water + ethanol + NaCl)					
MIAC					
Esteso et al. 1989 ⁷⁰	R	298.15	0.038	0.200 - 0.900	123
Mamontov et al. 2016 ⁷¹	R	288.15 - 318.15	0.055	0.100 - 0.400	117
VLE					
Yang et al. 1979 ⁴⁶	R	298.15	0.025	0.100 - 0.700	22
Meyer et al. 1991 91	Т	306.35 - 332.05	0.072	0.013 - 0.904	30
(water + ethanol + KCl)					
MIAC					
Mussini et al. 1995 ⁷²	R	298.15	0.038	0.200 - 0.400	24
VLE					
Yang et al. 1979 ⁴⁶	R	298.15	0.021	0.100 - 0.600	14
Sun 1996 92	R	352.2 - 356.1	0.029	0.378 - 0.714	10
(water + ethanol + CsCl)					
MIAC					
Mussini et al. 1995 ⁷²	Т	298.15	0.112	0.200 - 0.700	59
(water + ethanol + NaF)					
MIAC					
Hernandez-Luis et al. 2003 ⁸⁶	Т	298.15	0.011	0.100 - 0.800	89
(water + ethanol + LiBr)					
VLE					
Sun 1996 92	U	352.51 - 356.07	0.050	0.378 - 0.714	12
(water + ethanol + NaBr)					
MIAC					
Gonzalez-Diaz et al. 1995 93	U	298.15	0.027	0.200 - 0.998	100
Han & Pan 1993 88	U	298.15	0.083	0.100 - 0.900	120
VLE					
Sun 1996 92	U	351.46 - 356.15	0.049	0.378 - 0.856	22
(water + ethanol + KBr)					
VLE					

Reference	Group	<i>T</i> (K)	$Max(x_{ion})$	W^0_{alc}	$N_{\rm DP}$
Burns & Furter 1976 ⁹⁴	Т	354.65 - 356.75	0.090	0.399 - 0.536	35
(water + ethanol + CsBr)					
MIAC					
Du et al. 2012 95	U	298.15	0.019	0.100 - 0.300	57
(water + ethanol + NaI)					
VLE					
Yamamoto et al. 1995 ⁹⁶	Т	298.15	0.052	0.140 - 0.968	9
(water + ethanol + KI)					
VLE					
Sun 1996 92	U	351.63 - 356.05	0.038	0.533	15
Burns & Furter 1979 97	Т	355.25 - 357.15	0.106	0.384 - 0.856	21

500 Table 5. Overall ADs of the E-NRTL model from the experimental datasets



501



505 **5 Benchmark database**

506 In this section, the benchmark databases of the aqueous and mixed-solvent electrolyte 507 solutions are presented.

508 **5.1 Database for aqueous electrolyte solutions**

509 Compared to the data status of the mixed-solvent electrolyte solutions, that of the aqueous electrolyte binary mixtures is much more extensive, which facilitates selection of the datasets that 510 511 are validated in higher accuracy. Thus, only the selected datasets are shown here. The accepted MIAC and VLE datasets of the (water + salt) mixtures are summarized in the Supporting 512 513 Information. The maximum x_{ion} column in the summary table shows the maximum value in the entire dataset. However, only the part that is smaller than $x_{ion} = 0.06$ is used in the regression and 514 comparison, because the E-NRTL model deviates significantly from experimental values at large 515 516 salt composition, as discussed in Section 3.3.

517 **5.2 Database for mixed-solvent electrolyte solutions**

518 The MIAC and VLE datasets of the (water + alcohol + salt) mixed-solvent electrolyte 519 solutions are summarized in Table 4. Compared to the data status of the aqueous electrolyte 520 solutions, that of the mixed-solvent electrolyte solutions is much less extensive. In some cases, there is only one dataset, or there are datasets that are contradictory to each other, so that they 521 522 cannot be verified. Therefore, datasets are marked as "R" (recommended), "T" (tentative), and "U" 523 (unknown). Group **R** denotes datasets that have been verified against other datasets with deviations 524 within a few percent, e.g., the datasets of (water + methanol + NaCl). In many cases, such 525 verifications are not possible, because data are scarce or contradictory between difference sources. 526 In these cases, when there is only one dataset for the mixture, group T denotes datasets that are 527 considered reliable by us, based on data trends and ranges, while group U denotes datasets that are 528 considered unreliable by us. When there are multiple datasets available for the mixture, group T denotes datasets that deviate from each other but not by too much, while group U denotes datasets 529 530 that deviate from each other by a lot so that it is not possible to decide which one or ones are reliable. As discussed in Section 2.1, experimental literature does not report the combined 531 532 uncertainty of MIAC. For group R datasets, data from different sources agree with each other 533 within a few percent; in most cases, one can expect the data to be very accurate. For group T datasets, discrepancy between data of different sources exceed 10% at boundaries of data ranges. 534 For group U datasets, as no comparison can be made between data of different sources, accuracy 535 cannot be estimated. The data tables are provided in the Supporting Information. Group **R** and **T** 536 537 datasets are provided in the same file, while group U datasets are provided in a separate file. For datasets that are marked as T and U, reasons are given in the following text in this section. These
explanations are to be read together with the dataset summary in Table 4.

For (water + methanol + LiCl), although there are a few datasets, they do not agree with each 540 other. The Hu et al. dataset ⁷⁶ is smaller than the Harned dataset ⁷⁴ by a few percent, while the 541 datasets from the Mussini group ^{45,75} is larger by a few percent. A maximum deviation that is 542 approximately 10% is observed in the only VLE dataset ⁷⁷ as it is regressed along with the MIAC 543 datasets. Therefore, these datasets are all marked as **T**. For this mixture, there is another MIAC 544 545 dataset ⁹⁸, which also covers a few other mixtures in Table 4. However, it is rejected because that scattering is up to a few percent within their own datasets for a few mixtures, and that these datasets 546 547 do not agree with the **R** datasets that are available for some of these mixtures. There is another VLE dataset ⁸⁷. However, after regression, large deviations are observed for vapor phase molar 548 fraction, which indicates that it is not reliable. Furthermore, datasets from the same reference are 549 found to deviate significantly from the **R** datasets of a few other mixtures. Therefore, they are not 550 551 included in the table.

For (water + methanol + NaCl), apart from the datasets summarized in Table 4, there are a few other datasets for MIAC and VLE $^{74,98-102}$. However, the **R** datasets agree very well with each other. Therefore, these datasets are not included in the table. For MIAC, datasets from references 98,99 deviate significantly from the **R** datasets, while datasets from references 74,100,101 deviate from the **R** datasets by only a few percent. For VLE, there is another dataset 102 . However, the vapor phase molar fraction deviates significantly from the **R** datasets.

For (water + methanol + KCl), the Harned ⁷⁴ dataset is marked as **T** because the Basili et al. ⁶⁹ dataset agrees very well when regressed together with the VLE datasets ^{68,81}, while it deviates by a few percent. Because the Basili et al. ⁶⁹ MIAC dataset covers larger composition range, the Harned ⁷⁴ dataset is not included in parameter regression. There are another MIAC dataset ⁹⁸ and another VLE dataset ⁸⁷. However, they are not included in this table because they deviate from the **R** and **T** datasets significantly.

For (water + methanol + CsCl), the three available MIAC datasets ^{83–85} deviate significantly from each other. Therefore, no conclusion can be drawn and they are marked as **U**. For (water + methanol + NaF), the Boone et al. VLE dataset 87 is marked as U because that it cannot be regressed to an accuracy within 10% even though NaF has a very low salt solubility, and that datasets from the same reference are found to deviate significantly from the **R** datasets of a few other mixtures.

For (water + methanol + NaBr), the MIAC dataset ⁸⁸ and Boone et al. VLE dataset ⁸⁷ are 570 571 marked as U, because they deviate significantly when regressed together. Plotted on the graph (as shown in the Supporting Information), the MIAC data are clustered unreasonably: 40% ethanol 572 573 weight fraction (salt-free basis) data are clustered with 60% data; while large gaps are present between the 20% and 40% data, and between the 60% and 80% data. Such behavior is not 574 575 successfully correlated with the E-NRTL model. Thus, verifications against the U MIAC datasets cannot conclude on the reliability of the Yang and Lee VLE dataset ⁶⁸. However, datasets from the 576 same reference have been verified for other mixtures. In addition, it is represented well with the E-577 NRTL model. Furthermore, the Yang and Lee dataset only covers small alcohol composition, and 578 579 thus can be considered to be verified against the accepted (water + NaBr) datasets. Therefore, it is 580 marked as **T**.

For (water + methanol + CsI), the Falciola et al. ⁸⁵ dataset is the only available dataset, and thus cannot be verified. In addition, the Falciola et al. dataset present a strange behavior at high alcohol composition (75% weight fraction on the salt-free basis): at 0.018 ion-based molar fraction (approximately 0.5 M), MIAC increases to as large as 5. Such behavior is not observed in the (water + methanol + CsCl) and (water + methanol + CsBr) datasets and the lower-alcohol-composition part of the (water + methanol + CsI) dataset from the same reference, and cannot be represented with the E-NRTL model. Therefore, it is marked as U.

For (water + methanol + CsBr), the Falciola et al. ⁸⁵ dataset is also marked as **U**. There is another dataset for this mixture by Du et al. ⁹⁵. However, MIAC decreases to very small values in the Du et al. ⁹⁵ dataset. Such behavior is not observed in any of the **R** and **T** datasets of the investigated mixtures. Therefore, it is not included in Table 4.

For (water + ethanol + LiCl), the two MIAC datasets ^{76,89} and the VLE dataset ⁹⁰ do not agree
with each other. The 5%-15% ethanol weight fraction (salt-free basis) data from the Hu et al. ⁷⁶
dataset is smaller than the 20% data from the Hernandez-Luis dataset ⁸⁹, which is against the overall

595 trend that MIAC is smaller for the data at larger alcohol composition. Regression with the two MIAC datasets together with the VLE dataset does not result in any preference between them. In 596 addition, there are two other VLE datasets ^{92,103}. However, they present vapor phase molar fraction 597 that deviates significantly after regression, which is very unlikely for reliable data. Therefore, they 598 are not included in the table. A dataset from reference ¹⁰³ is also rejected after comparison with the 599 recommended datasets for (water + ethanol + NaCl). However, datasets from reference ⁹² agrees 600 well with the other **R** datasets for (water + ethanol + KCl) (marked as **R**), (water + ethanol + KI) 601 (marked as **R**), and (water + ethanol + NaBr) (marked as **T** because the quality of other datasets of 602 603 this mixture is not as good as that of the KCl and KI mixtures). In the meantime, being the only dataset for (water + ethanol + LiBr), a dataset from the same reference cannot be represented 604 605 accurately after regression, and is marked as **U**.

For (water + ethanol + NaCl), in addition to the VLE dataset from reference 103 , there is an additional MIAC dataset 104 that is not included in Table 4, because it deviates significantly from the **R** datasets. The Meyer et al. 91 dataset is marked as **T**, because its vapor phase molar fraction deviation is slightly larger than 10% after regression, while the other datasets agree very well with the model. Despite this, it is included in the table because it is the only VLE dataset at temperatures other than 298.15 K.

For (water + ethanol + CsCl), the Mussini et al. 72 dataset is marked as **T**. It is represented 612 very well with the E-NRTL model. There is another MIAC dataset for this mixture by Hu et al.⁸³. 613 614 However, the data decreases to very small values at smaller than 0.008 ion-based molar fraction. Such behavior is not observed in any other investigated mixtures. A dataset from reference ⁸³ is 615 616 among the three MIAC datasets of (water + methanol + CsCl) that are contradictory to each other. In addition, a dataset from the same reference is rejected in the evaluation for the (water + CsCl) 617 618 datasets, although not because of wrong trend, but rather only because of relatively larger deviations compared to the accepted datasets. Therefore, the Hu et al.⁸³ dataset is not included in 619 620 Table 4.

For (water + ethanol + NaF), the Hernandez-Luis et al. dataset ⁸⁶ is the only available dataset.
Although it is not verified against other datasets, it is unlikely to be far off because of the very
small solubility of NaF. In addition, it is very well represented with the E-NRTL model. Therefore,
it is marked as T.

For (water + ethanol + NaBr), in the same range as the Gonzalez-Dias et al. dataset 93 , the Han 625 and Pan dataset ⁸⁸ is slightly smaller. Furthermore, at larger salt composition, the upward curvature 626 that seems quite artificial cannot be represented with the E-NRTL model. There is no dataset to 627 verify the Han and Pan dataset, considering that the only VLE dataset is from a reference ⁹² in 628 which datasets have been evaluated to be unreliable for a few other mixtures. In addition, a dataset 629 is marked as U from the reference ⁸⁸ for (water + methanol + NaBr). Therefore, the VLE dataset 630 are marked as T, while the MIAC datasets are both marked as U. However, because no dataset of 631 better quality is available, all three datasets have been used in the regression. 632

For (water + ethanol + KBr) and (water + ethanol + NaI), each mixture only has one VLE
dataset ^{94,96}. Thus, they are not verified, but are represented very well with the E-NRTL model.
Therefore, they are marked as T.

For (water + ethanol + CsBr), the only dataset is from reference 95 . It has the same problem as the (water + methanol + CsBr) from the same reference, presenting very small MIAC values that are not observed in any of the **R** and **T** datasets in the investigated mixtures. Therefore, it is marked as **U**.

For (water + ethanol + KI), there is another VLE dataset 105 that deviates significantly from the **T** dataset. It is not included in Table 4. The Sun 92 dataset deviates more compared to the **T** dataset after regression. Considering that datasets from the same reference also present larger deviations compared to the **R** datasets of other mixtures when more detailed comparisons are possible, the Sun 92 dataset is marked **U**. Therefore, the remaining VLE dataset cannot be verified, and is marked as **T**.

Figure 9 shows the data status of the (water + methanol/ethanol + alkali halide) ternary mixtures. As the datasets are evaluated, 13 of the 20 mixtures for which experimental data are available can be confirmed against data from other sources. From the perspectives of data quality and availability, we recommend that further experimental work be conducted on mixtures that only have U datasets (red), and on mixtures that have not been measured yet (white).



651

Figure 9. Data status of the ternary mixtures: (a) (water + methanol + alkali halide), (b) (water + ethanol + alkali halide). The green color denotes that there are **R** data for both MIAC and VLE. The light green
color denotes that there are **R** data for MIAC only. The blue color denotes that there are **T** data for MIAC.
The orange color denotes that there are **R** or **T** data for VLE only. The red color denotes that all available data are marked as **U**. The white color denotes that there are no data.

657 **6 Conclusions and perspectives**

In this work, a benchmark database for (water + methanol/ethanol + alkali halide) mixedsolvent electrolyte solutions is presented. A consistent E-NRTL model that satisfies the Gibbs-Duhem equation is utilized for data analysis, reconciling solvent and ion composition derivatives of the excess Gibbs energy, and thus reconciling MIAC and VLE. Major conclusions are:

Available experimental data of MIAC and VLE are comprehensively collected and critically
 evaluated for the 61 datasets of 20 solutions. The resulting benchmark database covers 1413
 data records from 32 datasets for 13 solutions. Evaluated datasets that are used in the
 parameterization of the relevant water-salt pair are also presented.

666 2. The consistent E-NRTL model is compared against the widely used original E-NRTL model.
667 Results and parameters indicate that the consistent model better reconciles MIAC and VLE,
668 especially the salting out effect of the co-solvent (alcohol) when it is highly diluted in water.

A consistency analysis is conducted within each solution. Impacts of the objective functions
and water-salt parameters are discussed, facilitating the parameterization and data analysis.
Recommended parameters are obtained based on the obtained benchmark database.

672 Considering that alcohol and monovalent strong salts represent the simplest components in 673 mixed-solvent electrolyte solutions, we propose that the database be used as a benchmark for model 674 development and evaluation, as thermodynamic property models are extended to mixed-solvent 675 electrolyte solutions.

676 Nomenclature

677	A_{arphi}	One third of the Debye-Huckel limiting slope
678	AD	Average deviation
679	E-NRTL	Electrolyte non-random two-liquid
680	G^{E}	Excess Gibbs energy
681	I_x	Ionic strength
682	k	Boltzmann constant
683	m	Molality
684	М	Molar mass
685	MD	Maximum deviation
686	MIAC	Mean ionic activity coefficient
687	n	Molar amount
688	N _A	Avogadro number

689	OF	Objective function
690	OF-M	OF = MIAC
691	OF-MV	OF = MIAC + VLE
692	OF-V	OF = VLE
693	p	Pressure
694	PDH	Pitzer-Debye-Huckel
695	$Q_{ m e}$	Elementary charge
696	r	Born radius
697		Pauling radius
698	R	Universal gas constant
699	R	Recommended datasets
700	SLE	Solid-liquid equilibrium
701	Т	Temperature
702	Τ	Tentative datasets
703	U	Uncertain datasets
704	VLE	Vapor-liquid equilibrium
705	w^0	Weight fraction on the salt-free basis
706	x	Molar fraction (in general or in the liquid phase)
707	x^0	Composition in the original unit
708		Molar fraction on the salt-free basis
709	у	Molar fraction in the vapor phase

710	Ζ	Ionic charge
711	Greek letters	
712	α	Parameter in the NRTL term
713	γ	Activity coefficient
714	$\Delta g_{ m trans}$	Molar Gibbs energy of transfer
715	$\Delta_{ m f} g$	Standard Gibbs energy of formation
716	Δp	Vapor pressure depression
717	ε	Relative permittivity
718	\mathcal{E}_0	Dielectric constant in the vacuum
719	ν	Stoichiometric coefficient
720	ρ	Density
721	τ	Parameter in the NRTL term
722	Subscript	
723	±	Mean ionic
724	a	Anion
725	(a)	Reference state at infinite dilution in water
726	с	Cation
727	m	Molecule
728	S	Solvent mixture
729	(s)	Solid phase
730	W	Water

731	Superso	cript
732	*	Rational unsymmetrical
733	∞	Infinite dilution
734	m	Molality
735	c	Molarity
736	sat	Saturation state of pure solvent
737	Suppo	rting Information
738	Su	oporting Information are provided for the following contents:
739	-	Analytical expressions for $\left(\frac{\partial A_{\varphi}}{\partial n_i}\right)_{T,p,n_{j(j\neq i)}}$ and the other derivatives over n_i .
740	-	Solvent activity coefficients of (water + methanol + KCl) and (water + ethanol +
741		NaCl/KCl).
742	-	Parameters for solvent component density and relative permittivity.
743	-	Parameters of the consistent and original E-NRTL models for (water + methanol/ethanol
744		+ NaCl/KCl) with OF-M and OF-MV.
745	-	Deviations of MIAC calculated with the E-NRTL model with OF-MV and parameters
746		regressed in the region up to $x_{ion} = 0.06$ and in the entire x_{ion} range.
747	-	Average and maximum deviations of the e-NRTL model from the experimental datasets.
748	-	Summary of accepted MIAC and VLE datasets for aqueous electrolyte solutions.
749	-	Benchmark database files for the mixed-solvent electrolyte solutions and relevant
750		aqueous electrolyte solutions.
751	-	MIAC results for the mixed-solvent electrolyte solutions.
752	The	e information is available free of charge via the Internet at http://pubs.acs.org/.

753 Notes

754 The authors declare no competing financial interest.

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759 **Reference:**

760 (1) Bui, M.; Adjiman, C. S.; Bardow, A.; Anthony, E. J.; Boston, A.; Brown, S.; Fennell, P.

- 761 S.; Fuss, S.; Galindo, A.; Hackett, L. A.; Hallett, J. P.; Herzog, H. J.; Jackson, G.; Kemper,
- J.; Krevor, S.; Maitland, G. C.; Matuszewski, M.; Metcalfe, I. S.; Petit, C.; Puxty, G.;
- Reimer, J.; Reiner, D. M.; Rubin, E. S.; Scott, S. A.; Shah, N.; Smit, B.; Trusler, J. P. M.;
- Webley, P.; Wilcox, J.; Mac Dowell, N. Carbon Capture and Storage (CCS): The Way
- 765 Forward. *Energy Environ. Sci.* **2018**, *11* (5), 1062–1176.
- 766 https://doi.org/10.1039/c7ee02342a.
- 767 (2) Gholami, F.; Tomas, M.; Gholami, Z.; Vakili, M. Technologies for the Nitrogen Oxides
 768 Reduction from Flue Gas: A Review. *Sci. Total Environ.* 2020, *714*, 136712.
- 769 https://doi.org/10.1016/j.scitotenv.2020.136712.
- (3) Darre, N. C.; Toor, G. S. Desalination of Water: A Review. *Curr. Pollut. Reports* 2018, 4
 (2), 104–111. https://doi.org/10.1007/s40726-018-0085-9.
- (4) Løge, I. A.; Bentzon, J. R.; Klingaa, C. G.; Walther, J. H.; Anabaraonye, B. U.; Fosbøl, P.
- L. Scale Attachment and Detachment: The Role of Hydrodynamics and Surface
- 774 Morphology. *Chem. Eng. J.* **2021**, *430* (P2), 132583.
- 775 https://doi.org/10.1016/j.cej.2021.132583.
- 776 (5) Olasunkanmi, L. O. Corrosion: Favoured, yet Undesirable Its Kinetics and
- Thermodynamics. In *Corrosion*; IntechOpen: London, 2021; pp 1–17.
- 778 (6) Yu, Z.; Wang, H.; Kong, X.; Huang, W.; Tsao, Y.; Mackanic, D. G.; Wang, K.; Wang, X.;
- Huang, W.; Choudhury, S.; Zheng, Y.; Amanchukwu, C. V.; Hung, S. T.; Ma, Y.; Lomeli,
- 780 E. G.; Qin, J.; Cui, Y.; Bao, Z. Molecular Design for Electrolyte Solvents Enabling
- 781 Energy-Dense and Long-Cycling Lithium Metal Batteries. *Nat. Energy* **2020**, *5* (7), 526–

- 782 533. https://doi.org/10.1038/s41560-020-0634-5.
- Molla, G. S.; Freitag, M. F.; Stocks, S. M.; Nielsen, K. T.; Sin, G. Solubility Prediction of
 Different Forms of Pharmaceuticals in Single and Mixed Solvents Using Symmetric
- 785Electrolyte Nonrandom Two-Liquid Segment Activity Coefficient Model. Ind. Eng. Chem.
- 786 *Res.* **2019**, 58 (10), 4267–4276. https://doi.org/10.1021/acs.iecr.8b04268.
- 787 (8) Bazyleva, A.; Abildskov, J.; Anderko, A.; Baudouin, O.; Chernyak, Y.; de Hemptinne, J.
- 788 C.; Diky, V.; Dohrn, R.; Elliott, J. R.; Jacquemin, J.; Jaubert, J. N.; Joback, K. G.; Kattner,
- U. R.; Kontogeorgis, G. M.; Loria, H.; Mathias, P. M.; O'Connell, J. P.; Schröer, W.;
- 790 Smith, G. J.; Soto, A.; Wang, S.; Weir, R. D. Good Reporting Practice for Thermophysical
- and Thermochemical Property Measurements (IUPAC Technical Report). *Pure Appl.*

792 *Chem.* **2021**, *93* (2), 253–272. https://doi.org/10.1515/pac-2020-0403.

- (9) May, P. M.; Rowland, D. Correction to Thermodynamic Modeling of Aqueous Electrolyte
- 794Systems: Current Status (Journal of Chemical & Engineering Data (2017) 62:9 (2481-
- 795
 2495) DOI: 10.1021/Acs.Jced.6b01055). J. Chem. Eng. Data 2019, 64 (7), 3212.
- 796 https://doi.org/10.1021/acs.jced.9b00468.
- 797 (10) Kontogeorgis, G. M.; Dohrn, R.; Economou, I. G.; De Hemptinne, J. C.; Kate, A.;
- 798 Kuitunen, S.; Mooijer, M.; Zilnik, L. F.; Vesovic, V. Industrial Requirements for
- Thermodynamic and Transport Properties: 2020. Ind. Eng. Chem. Res. 2021, 60 (13),
- 4987–5013. https://doi.org/10.1021/acs.iecr.0c05356.
- Kontogeorgis, G. M.; Maribo-Mogensen, B.; Thomsen, K. The Debye-Hückel Theory and
 Its Importance in Modeling Electrolyte Solutions. *Fluid Phase Equilib.* 2018, 462, 130–
 152. https://doi.org/10.1016/j.fluid.2018.01.004.
- 804 (12) Simonin, J. P. On the "Born" Term Used in Thermodynamic Models for Electrolytes. J.
 805 *Chem. Phys.* 2019, *150*, 244503. https://doi.org/10.1063/1.5096598.

806 (13) Thomsen, K. Electrolyte Solutions: Thermodynamics, Crystallization, Separation

- 807 *Methods*; Technical University of Denmark: Kgs. Lyngby, 2009.
- 808 https://doi.org/10.11581/dtu:00000073.
- 809 (14) Vaque Aura, S.; Roa Pinto, J. S.; Ferrando, N.; De Hemptinne, J. C.; Ten Kate, A.;
- 810 Kuitunen, S.; Diamantonis, N.; Gerlach, T.; Heilig, M.; Becker, G.; Brehelin, M. Data
- 811 Analysis for Electrolyte Systems: A Method Illustrated on Alkali Halides in Water. J.
- 812 *Chem. Eng. Data* **2021**, *66* (8), 2976–2990. https://doi.org/10.1021/acs.jced.1c00105.

- 813 (15) Raatschen, W.; Harvey, A. H.; Prausnitz, J. M. Equation of State for Solutions of
 814 Electrolytes in Mixed Solvents. *Fluid Phase Equilib.* 1987, *38* (1–2), 19–38.
- 815 https://doi.org/10.1016/0378-3812(87)90002-1.
- 816 (16) Zuo, Y. X.; Fürst, W. Use of an Electrolyte Equation of State for the Calculation of Vapor-
- 817 Liquid Equilibria and Mean Activity Coefficients in Mixed Solvent Electrolyte Systems.
- 818 Fluid Phase Equilib. **1998**, 150 (151), 267–275. https://doi.org/10.1016/s0378-
- 819 3812(98)00326-4.
- 820 (17) Simon, H. G.; Kistenmacher, H.; Prausnitz, J. M.; Vortmeyer, D. An Equation of State for
 821 Systems Containing Electrolytes and Nonelectrolytes. *Chem. Eng. Process.* 1991, *29* (3),
 822 139–146. https://doi.org/10.1016/0255-2701(91)85013-E.
- (18) Zuo, J. Y.; Zhang, D.; Fürst, W. Predicting LLE in Mixed-Solvent Electrolyte Systems by
 an Electrolyte EOS. *AIChE J.* 2000, *46* (11), 2318–2329.
- 825 https://doi.org/10.1002/aic.690461122.
- (19) Zuo, J. Y.; Zhang, D.; Fürst, W. Extension of the Electrolyte EOS of Furst and Renon to
 Mixed Solvent Electrolyte Systems. *Fluid Phase Equilib.* 2000, *175* (1–2), 285–310.
 https://doi.org/10.1016/S0378-3812(00)00463-5.
- 829 (20) Held, C.; Reschke, T.; Mohammad, S.; Luza, A.; Sadowski, G. EPC-SAFT Revised.
- 830 *Chem. Eng. Res. Des.* **2014**, *92* (12), 2884–2897.
- 831 https://doi.org/10.1016/j.cherd.2014.05.017.
- 832 (21) Schreckenberg, J. M. A.; Dufal, S.; Haslam, A. J.; Adjiman, C. S.; Jackson, G.; Galindo,
- A. Modelling of the Thermodynamic and Solvation Properties of Electrolyte Solutions
- 834 with the Statistical Associating Fluid Theory for Potentials of Variable Range. *Mol. Phys.*

2014, *112* (17), 2339–2364. https://doi.org/10.1080/00268976.2014.910316.

- 836 (22) Maribo-Mogensen, B.; Thomsen, K.; Kontogeorgis, G. M. An Electrolyte CPA Equation
- 837 of State for Mixed Solvent Electrolytes. *AIChE J.* **2015**, *61* (9), 2933–2950.
- 838 https://doi.org/10.1002/aic.14829.
- 839 (23) Das, G.; dos Ramos, M. C.; McCabe, C. Predicting the Thermodynamic Properties of
- Experimental Mixed-Solvent Electrolyte Systems Using the SAFT-VR+DE Equation of
 State. *Fluid Phase Equilib.* 2018, 460, 105–118.
- 842 https://doi.org/10.1016/j.fluid.2017.11.017.
- 843 (24) Ahmed, S.; Ferrando, N.; de Hemptinne, J. C.; Simonin, J. P.; Bernard, O.; Baudouin, O.

- 844 Modeling of Mixed-Solvent Electrolyte Systems. *Fluid Phase Equilib.* **2018**, *459*, 138–
- 845 157. https://doi.org/10.1016/j.fluid.2017.12.002.
- 846 (25) Neau, E.; Nicolas, C.; Avaullée, L. Extension of the Group Contribution NRTL-PRA EoS
 847 for the Modeling of Mixtures Containing Light Gases and Alcohols with Water and Salts.
- 848 *Fluid Phase Equilib.* **2018**, *458*, 194–210. https://doi.org/10.1016/j.fluid.2017.09.028.
- 849 (26) Mock, B.; Evans, L. B.; Chen, C. -C. Thermodynamic Representation of Phase Equilibria
 850 of Mixed-solvent Electrolyte Systems. *AIChE J.* 1986, *32* (10), 1655–1664.
- 851 https://doi.org/10.1002/aic.690321009.
- (27) Chen, C. C.; Song, Y. Generalized Electrolyte-NRTL Model for Mixed-Solvent
 Electrolyte Systems. *AIChE J.* 2004, *50* (8), 1928–1941. https://doi.org/10.1002/aic.10151.
- 854 (28) Macedo, E. A.; Skovborg, P.; Rasmussen, P. Calculation of Phase Equilibria for Solutions
- 855 of Strong Electrolytes in Solvent-Water Mixtures. *Chem. Eng. Sci.* 1990, 45 (4), 875–882.
 856 https://doi.org/10.1016/0009-2509(90)85009-3.
- 857 (29) Sander, B.; Fredenslund, A.; Rasmussen, P. Calculation of Vapour-Liquid Equilibria in
 858 Mixed Solvent/Salt Systems Using an Extended UNIQUAC Equation. *Chem. Eng. Sci.*859 1986, 41 (5), 1171–1183. https://doi.org/10.1016/0009-2509(86)87090-7.
- 860 (30) Zerres, H.; Prausnitz, J. M. Thermodynamics of Phase Equilibria in Aqueous-organic
 861 Systems with Salt. *AIChE J.* 1994, 40 (4), 676–691.
- 862 https://doi.org/10.1002/aic.690400411.
- (31) Van Bochove, G. H.; Krooshof, G. J. P.; De Loos, T. W. Modelling of Liquid-Liquid
 Equilibria of Mixed Solvent Electrolyte Systems Using the Extended Electrolyte NRTL. *Fluid Phase Equilib.* 2000, *171* (1–2), 45–58. https://doi.org/10.1016/S03783812(00)00347-2.
- 867 (32) Thomsen, K. Modeling Electrolyte Solutions with the Extended Universal Quasichemical
 868 (UNIQUAC) Model. *Pure Appl. Chem.* 2005, 77 (3), 531–542.
- 869 https://doi.org/10.1351/pac200577030531.
- (33) Djamali, E.; Kan, A. T.; Tomson, M. B. A Priori Prediction of Thermodynamic Properties
 of Electrolytes in Mixed Aqueous-Organic Solvents to Extreme Temperatures. *J. Phys.*
- 872 *Chem. B* **2012**, *116* (30), 9033–9042. https://doi.org/10.1021/jp301857m.
- 873 (34) Djamali, E.; Tomson, M. B.; Chapman, W. G. Thermodynamic Properties and Solubility
 874 of Sodium and Potassium Chloride in Ethane-1,2-Diol/Water Mixed Solvent Systems to

- 875 High Temperatures. J. Chem. Eng. Data 2017, 62 (4), 1326–1334.
- 876 https://doi.org/10.1021/acs.jced.6b00842.
- 877 (35) Hála, E. Vapor-Liquid Equilibria of Strong Electrolytes in Systems Containing Mixed
 878 Solvent. *Fluid Phase Equilib.* 1983, *13*, 311–319.
- 879 https://doi.org/https://doi.org/10.1016/0378-3812(83)80102-2.
- 880 (36) Iliuta, M. C.; Thomsen, K.; Rasmussen, P. Extended UNIQUAC Model for Correlation
- and Prediction of Vapour-Liquid-Solid Equilibria in Aqueous Salt Systems Containing
- Non-Electrolytes. Part A. Methanol-Water-Salt Systems. *Chem. Eng. Sci.* 2000, *55* (14),
 2673–2686. https://doi.org/10.1016/S0009-2509(99)00534-5.
- 884 (37) Thomsen, K.; Iliuta, M. C.; Rasmussen, P. Extended UNIQUAC Model for Correlation

and Prediction of Vapor-Liquid-Liquid-Solid Equilibria in Aqueous Salt Systems

- 886 Containing Non-Electrolytes. Part B. Alcohol (Ethanol, Propanols, Butanols)-Water-Salt
- 887 Systems. Chem. Eng. Sci. 2004, 59 (17), 3631–3647.
- 888 https://doi.org/10.1016/j.ces.2004.05.024.
- (38) Fosbøl, P. L.; Thomsen, K.; Stenby, E. H. Modeling of the Mixed Solvent Electrolyte
 System CO2-Na 2CO3-NaHCO3-Monoethylene Glycol-Water. *Ind. Eng. Chem. Res.*2009, 48 (9), 4565–4578. https://doi.org/10.1021/ie801168e.
- (39) Guggenheim, E. A. L. The Specific Thermodynamic Properties of Aqueous Solutions of
 Strong Electrolytes. *London, Edinburgh, Dublin Philos. Mag. J. Sci.* 1935, *19* (127), 588–
 643. https://doi.org/10.1080/14786443508561403.
- (40) Christensen, C.; Sander, B.; Fredenslund, A.; Rasmussen, P. Towards the Extensionof
 UNIFAC to Mixtures with Electrolytes. *Fluid Phase Equilib.* 1983, *13* (C), 297–309.
 https://doi.org/10.1016/0378-3812(83)80101-0.
- 898 (41) Pitzer, K. S. *Activity Coefficients in Electrolyte Solutions*, 2nd Editio.; CRC Press: Boca
 899 Raton Ann Arbor Boston London, 1991.
- 900 (42) Robinson, R. A.; Sinclair, D. A. The Activity Coefficients of the Alkali Chlorides and of
 901 Lithium Iodide in Aqueous Solution from Vapor Pressure Measurements. *J. Am. Chem.*902 Soc. 1934, 56 (9), 1830–1835. https://doi.org/10.1021/ja01324a003.
- 903 (43) Robinson, R. A. The Osmotic and Activity Coefficient Data of Some Aqueous Salt
- Solutions from Vapor Pressure Measurements. J. Am. Chem. Soc. 1937, 59 (1), 84–90.
 https://doi.org/10.1021/ja01280a019.
 - 45

- 906 (44) Robinson, R. A. The Activity Coefficients of Some Alkali Halides at 25.°. *Trans. Faraday*907 Soc. 1939, 35, 1217–1220.
- 908 (45) Mussini, P. R.; Mussini, T.; Sala, B. Thermodynamics of the Cell {Li-Amalgam | LiX (m)
- 909 $| AgX | Ag \} (X = Cl,Br)$ and Medium Effects upon LiX in (Acetonitrile + Water), (1,4-
- 910 Dioxane + Water), and (Methanol + Water) Solvent Mixtures with Related Solvation
- 911 Parameters. J. Chem. Thermodyn. 2000, 32 (5), 597–616.
- 912 https://doi.org/10.1006/jcht.1999.0622.
- 913 (46) Yang, R.; Demirgian, J.; Solsky, J. F.; Kikta, E. J.; Marinsky, J. A. Mean Molal Activity of
 914 Sodium Chloride, Potassium Chloride, and Cesium Chloride in Ethanol-Water Mixtures. *J.*
- 915 *Phys. Chem.* **1979**, *83* (21), 2752–2761. https://doi.org/10.1021/j100484a013.
- 916 (47) Westhaus, U. DETHERM. DECHEMA e.V.: Frankfurt am Main 2019.
- 917 (48) Thomsen, K. Data Bank for Electrolyte Solutions. Technical University of Denmark: Kgs.
 918 Lyngby 2020.
- 919 (49) Advanced Tools for Optimization and Uncertainty Treatment. IFP Energies nouvelles:
 920 Rueil-Malmaison 2019.
- 921 (50) Span, R.; Wagner, W. A New Equation of State for Carbon Dioxide Covering the Fluid
 922 Region from the Triple-Point Temperature to 1100 K at Pressures up to 800 MPa. *J. Phys.*

923 *Chem. Ref. Data* **1996**, 25 (6), 1509–1596. https://doi.org/10.1063/1.555991.

- 924 (51) Chen, C.; Britt, H. I.; Boston, J. F.; Evans, L. B. Local Composition Model for Excess
 925 Gibbs Energy of Electrolyte Systems. Part I: Single Solvent, Single Completely
- 926 Dissociated Electrolyte Systems. *AIChE J.* **1982**, *28* (4), 588–596.
- 927 (52) Chen, C. -C; Evans, L. B. A Local Composition Model for the Excess Gibbs Energy of
 928 Aqueous Electrolyte Systems. *AIChE J.* 1986, *32* (3), 444–454.
- 929 https://doi.org/10.1002/aic.690320311.

930 (53) Chen, C. C. Toward Development of Activity Coefficient Models for Process and Product

- 931 Design of Complex Chemical Systems. *Fluid Phase Equilib.* **2006**, *241* (1–2), 103–112.
- 932 https://doi.org/10.1016/j.fluid.2006.01.006.
- 933 (54) Saravi, S. H.; Ravichandran, A.; Khare, R.; Chen, C. C. Bridging Two-Liquid Theory with
 934 Molecular Simulations for Electrolytes: An Investigation of Aqueous NaCl Solution.
- 935 *AIChE J.* **2019**, *65* (4), 1315–1324. https://doi.org/10.1002/aic.16521.
- 936 (55) Renon, H.; Prausnitz, J. M. Local Compositions in Thermodynamic Excess Functions for

- 937 Liquid Mixtures. *AIChE J.* **1968**, *14* (1), 135–144. https://doi.org/10.1002/aic.690140124.
- 938 (56) Pitzer, K. S. Electrolytes. From Dilute Solutions to Fused Salts. J. Am. Chem. Soc. 1980,
 939 102 (9), 2902–2906. https://doi.org/10.1021/ja00529a006.
- 940 (57) Liu, Y.; Watanasiri, S. Representation of Liquid-Liquid Equilibrium of Mixed-Solvent
 941 Electrolyte Systems Using the Extended Electrolyte NRTL Model. *Fluid Phase Equilib*.
- 942 **1996**, *116* (1–2), 193–200. https://doi.org/10.1016/0378-3812(95)02887-0.
- 943 (58) Robinson, R. A.; Stokes, R. H. *Electrolyte Solutions, 2nd Edition*; Dover Publications:
 944 Mineola, 2002.
- (59) Rashin, A. A.; Honig, B. Reevaluation of the Born Model of Ion Hydration. J. Phys. *Chem.* 1985, 89 (26), 5588–5593. https://doi.org/10.1021/j100272a006.
- 947 (60) Chang, C. K.; Lin, S. T. Extended Pitzer-Debye-Hückel Model for Long-Range
 948 Interactions in Ionic Liquids. *J. Chem. Eng. Data* 2020, 65 (3), 1019–1027.
- 949 https://doi.org/10.1021/acs.jced.9b00368.
- 950 (61) Song, Y.; Chen, C. C. Symmetric Electrolyte Nonrandom Two-Liquid Activity Coefficient
 951 Model. *Ind. Eng. Chem. Res.* 2009, 48 (16), 7788–7797.
- 952 https://doi.org/10.1021/ie9004578.
- 953 (62) Tsanas, C.; de Hemptinne, J.-C.; Mougin, P. Calculation of Phase and Chemical
 954 Equilibrium for Multiple Ion-Containing Phases Including Stability Analysis. *Chem. Eng.*

955 *Sci.* **2021**, 117174. https://doi.org/10.1016/j.ces.2021.117174.

- 956 (63) Mohr, P. J.; Newell, D. B.; Taylor, B. N. CODATA Recommended Values of the
 957 Fundamental Physical Constants: 2014. *J. Phys. Chem. Ref. Data* 2016, 45, 043102.
 958 https://doi.org/10.1063/1.4954402.
- 959 (64) Wilding, W. V.; Rowley, R. L.; Oscarson, J. L. DIPPR® Project 801 Evaluated Process
 960 Design Data. *Fluid Phase Equilib.* 1998, *150* (151), 413–420.
- 961 https://doi.org/10.1016/s0378-3812(98)00341-0.
- 962 (65) Raspo, I.; Neau, E. An Empirical Correlation for the Relative Permittivity of Liquids in a
- 963 Wide Temperature Range: Application to the Modeling of Electrolyte Systems with a
- 964 GE/EoS Approach. *Fluid Phase Equilib.* **2020**, *506*, 112371.
- 965 https://doi.org/10.1016/j.fluid.2019.112371.
- (66) Xu, Y.; Li, S.; Zhai, Q.; Jiang, Y.; Hu, M. Investigation on the Thermodynamic Properties
 of KCl/CsCl + NaCl + CH3OH + H2O Quaternary Systems at 298.15K. *J. Ind. Eng.*

- 968 *Chem.* **2014**, *20* (4), 2159–2165. https://doi.org/10.1016/j.jiec.2013.09.046.
- 969 (67) Basili, A.; Mussini, P. R.; Mussini, T.; Rondinini, S. Thermodynamics of the Cell:
- 970 {NaxHg1-X|NaCl(m)|AgCl|Ag} in (Methanol + Water) Solvent Mixtures. J. Chem.
 971 Thermodyn. 1996, 28 (8), 923–933. https://doi.org/10.1006/jcht.1996.0081.
- 972 (68) Yang, S. O.; Lee, C. S. Vapor-Liquid Equilibria of Water + Methanol in the Presence of
- 973 Mixed Salts. J. Chem. Eng. Data **1998**, 43 (4), 558–561.
- 974 https://doi.org/10.1021/je970286w.
- 975 (69) Basili, A.; Mussini, P. R.; Mussini, T.; Rondinini, S.; Vertova, A. Thermodynamics of the
 976 Amalgam Cell: {MexHg1-X|MeCl(m)| AgCl| Ag} (with Me = K, Rb) in (Methanol+water)
- 977 Solvent Mixtures. Berichte der Bunsengesellschaft für Phys. Chemie **1997**, 101 (5), 842–
- 978 846. https://doi.org/https://doi.org/10.1002/bbpc.19971010510.
- 979 (70) Esteso, M. A.; Gonzalez-Diaz, O. M.; Hernandez-Luis, F. F.; Fernandez-Merida, L.
- Activity Coefficients for NaCl in Ethanol-Water Mixtures at 25 C. J. Solution Chem. 1989,
 18 (3), 277–288.
- 982 (71) Mamontov, M. N.; Konstantinova, N. M.; Uspenskaya, I. A. Water-Ethanol-Sodium
 983 Chloride System: The Main Sources of Uncertainties in Thermodynamic Properties
- 984 Determined by Potentiometry. *Fluid Phase Equilib.* **2016**, *412*, 62–70.
- 985 https://doi.org/10.1016/j.fluid.2015.12.012.
- 986 (72) Mussini, P. R.; Mussini, T.; Perelli, A.; Rondinini, S.; Vertova, A. Thermodynamics of the
 987 Cell: {MexHg1-x|MeCl(m)|AgCl|Ag} (Me = Na,K,Cs) in (Ethanol + Water) Solvent
- 988 Mixtures. J. Chem. Thermodyn. **1995**, 27 (3), 245–251.
- 989 https://doi.org/https://doi.org/10.1006/jcht.1995.0022.
- (73) Yan, Y.; Chen, C. C. Thermodynamic Representation of the NaCl+Na2SO4+H2O System
 with Electrolyte NRTL Model. *Fluid Phase Equilib.* 2011, *306* (2), 149–161.
- 992 https://doi.org/10.1016/j.fluid.2011.03.023.
- 993 (74) Harned, H. S. A Rule for the Calculation of the Activity Coefficients of Salts in Organic
 994 Solvent-Water Mixtures. *J. Phys. Chem.* 1962, *66* (4), 589–591.
- 995 https://doi.org/10.1021/j100810a004.
- 996 (75) Basili, A.; Mussini, P. R.; Mussini, T.; Rondinini, S.; Sala, B.; Vertova, A. Transference
- 997 Numbers of Alkali Chlorides and Characterization of Salt Bridges for Use in
- 998 Methanol+water Mixed Solvents. J. Chem. Eng. Data **1999**, 44 (5), 1002–1008.

999

https://doi.org/10.1021/je9900979.

- 1000 (76) Hu, M.; Tang, J.; Li, S.; Xia, S.; Jiang, Y. Activity Coefficients of Lithium Chloride in
 1001 ROH/Water Mixed Solvent (R = Me, Et) Using the Electromotive Force Method at 298.15
 1002 K. J. Chem. Eng. Data 2008, 53 (2), 508–512. https://doi.org/10.1021/je700614h.
- 1003 (77) Broul, M.; Hlavatý, K.; Linek, J. Liquid-Vapour Equilibrium in Systems of Electrolytic
- 1004 Components. V. The System CH3OH-H2O-LiCl at 60 °C. *Collect. Czechoslov. Chem.* 1005 *Commun.* 1969, *34* (11), 3428–3435.
- 1006 (78) Yao, J.; Yan, W. D.; Xu, Y. J.; Han, S. J. Activity Coefficients for NaCl in MeOH + H2O
 1007 by Electromotive Force Measurements at 308.15 K and 318.15 K. J. Chem. Eng. Data
 1008 1999, 44 (3), 497–500. https://doi.org/10.1021/je970288g.
- 1009 (79) Hernández-Hernández, F.; Pérez-Villaseñor, F.; Hernández-Ruiz, V.; Iglesias-Silva, G. A.
 1010 Activity Coefficients of NaCl in H2O + MeOH + EtOH by Electromotive Force at 298.15
- 1011 K. J. Chem. Eng. Data 2007, 52 (3), 959–964. https://doi.org/10.1021/je600549h.
- 1012 (80) Yao, J.; Li, H.; Han, S. Vapor-Liquid Equilibrium Data for Methanol-Water-NaCl at
 1013 45 °C. *Fluid Phase Equilib.* 1999, *162* (1–2), 253–260. https://doi.org/10.1016/S03781014 3812(99)00204-6.
- 1015 (81) Johnson, A. I.; Furter, W. F. Salt Effect in Vapor-Liquid Equilibrium. *Can. J. Chem. Eng.*1016 **1960**, *38* (3), 78–87.
- 1017 (82) Zhang, J.; Gao, S.-Y.; Xia, S.-P.; Yao, Y. Study of Thermodynamic Properties of
 1018 Quaternary Mixture RbCl + Rb2SO 4 + CH 3OH + H2O by EMF Measurement at 298.15
 1019 K. *Fluid Phase Equilib.* 2004, 226 (1–2), 307–312.
- 1020 https://doi.org/10.1016/j.fluid.2004.10.010.
- 1021 (83) Hu, M.; Cui, R.; Li, S.; Jiang, Y.; Xia, S. P. Determination of Activity Coefficients for
- 1022 Cesium Chloride in Methanol-Water and Ethanol-Water Mixed Solvents by Electromotive
 1023 Force Measurements at 298.15 K. J. Chem. Eng. Data 2007, 52 (2), 357–362.
- 1024 https://doi.org/10.1021/je060278s.
- 1025 (84) Cui, R. F.; Hu, M. C.; Jin, L. H.; Li, S. N.; Jiang, Y. C.; Xia, S. P. Activity Coefficients of
 1026 Rubidium Chloride and Cesium Chloride in Methanol-Water Mixtures and a Comparative
- 1027 Study of Pitzer and Pitzer-Simonson-Clegg Models (298.15 K). *Fluid Phase Equilib.* 2007,
- 1028 251 (2), 137–144. https://doi.org/10.1016/j.fluid.2006.11.016.
- 1029 (85) Falciola, L.; Longoni, G.; Mussini, P. R.; Mussini, T. Thermodynamics of the Amalgam

- Cells {Cs-Amalgam|CsX (m)|AgX|Ag} (X = Cl, Br, I) and Primary Medium Effects in 1030 1031 (Methanol + Water), (Acetonitrile + Water), and (1,4-Dioxane + Water) Solvent Mixtures. 1032 J. Chem. Thermodyn. 2006, 38 (6), 788–798. https://doi.org/10.1016/j.jct.2005.08.014. Hernández-Luis, F.; Vázquez, M. V; Esteso, M. A. Activity Coefficients for NaF in 1033 (86)Methanol-Water and Ethanol-Water Mixtures at 25 °C. J. Mol. Lig. 2003, 108 (1), 283-1034 301. https://doi.org/https://doi.org/10.1016/S0167-7322(03)00187-9. 1035 1036 (87)BOONE, J. E.; ROUSSEAU, R. W.; SCHOENBORN, E. M. The Correlation of Vapor-Liquid Equilibrium Data for Salt-Containing Systems. In Thermodynamic Behavior of 1037 1038 Electrolytes in Mixed Solvents; Advances in Chemistry; AMERICAN CHEMICAL 1039 SOCIETY, 1976; Vol. 155, pp 36–52. https://doi.org/doi:10.1021/ba-1976-0155.ch004. 1040 (88) Han, S.; Pan, H. Thermodynamics of the Sodium Bromide-Methanol-Water and Sodium Bromide-Ethanol-Water Two Ternary Systems by the Measurements of Electromotive 1041 Force at 298.15K. Fluid Phase Equilib. 1993, 83 (C), 261–270. 1042 https://doi.org/10.1016/0378-3812(93)87029-Z. 1043 1044 (89) Hernández-Luis, F.; Galleguillos, H. R.; Graber, T. A.; Taboada, M. E. Activity Coefficients of LiCl in Ethanol - Water Mixtures at 298.15 K. Ind. Eng. Chem. Res. 2008, 1045 47 (6), 2056–2062. https://doi.org/10.1021/ie070704i. 1046 Shaw, R.; Butler, J. A. V. The Behavior of Electrolytes in Mixed Solvents. Part II. - The 1047 (90)Effect of Lithium Chloride on the Activities of Water and Alcohol in Mixed Solvents. 1048 Proc. R. Soc. London. Ser. A - Math. Phys. Sci. 1930, 129, 519–536. 1049 Meyer, T.; Poika, H. M.; Gmehling, J. Low-Pressure Isobarle Vapor-Liquid Equilibria of 1050 (91) Ethanol/Water Mixtures Containing Electrolytes. J. Chem. Eng. Data 1991, 36 (3), 340-1051 342. https://doi.org/10.1021/je00003a023. 1052 1053 (92)Sun, R. Molecular Thermodynamics of Salt Effect in Vapor-Liquid Equilibrium -Calculation of Isobaric VLE Salt Effect Parameters for Ethanol-Water-1-1 Type 1054 Electrolyte Systems. CIESC J. 1996, 47 (4), 401-409 (in Chinese). 1055 1056 (93) González-Díaz, O. M.; Fernández-Mérida, L.; Hernández-Luis, F.; Esteso, M. A. Activity 1057 Coefficients for NaBr in Ethanol-Water Mixtures at 25 °C. J. Solution Chem. 1995, 24 (6), 1058 551–563. https://doi.org/10.1007/BF00973206. 1059 (94)BURNS, J. A.; FURTER, W. F. Effects of Salts Having Large Organic Ions on Vapor-
 - 1060 Liquid Equilibrium. In Advances in Chemistry, Thermodynamic Behavior of Electrolytes

1061 *in Mixed Solvents*; Advances in Chemistry; AMERICAN CHEMICAL SOCIETY, 1976;

```
1062 Vol. 155, pp 99–127. https://doi.org/doi:10.1021/ba-1976-0155.ch008.
```

- 1063 (95) Du, Y.; Tang, J.; Li, S.; Zhai, Q.; Jiang, Y.; Hu, M. Activity Coefficients for CsBr +
- MeOH/EtOH + H2O Ternary Systems Using Potentiometric Measurements at 298.15 K. J. *Chem. Eng. Data* 2012, 57 (9), 2603–2609. https://doi.org/10.1021/je300671y.
- 1066 (96) Yamamoto, H.; Terano, T.; Yanagisawa, M.; Tokunaga, J.; Nishi, Y.; Tokunaga, J. Salt
- 1067 Effect of CACL2, NH4i and NAI on Vapor-Liquid Equilibrium of Ethanol + Water
- 1068 System at 298.15 K. Can. J. Chem. Eng. **1995**, 73 (5), 779–783.
- 1069 https://doi.org/https://doi.org/10.1002/cjce.5450730522.
- 1070 (97) Burns, J. A.; Furter, W. F. Salt Effect in Vapor-Liquid Equilibrium At Fixed Liquid
 1071 Composition. *Adv. Chem. Ser.* 1979, No. 177, 11–26. https://doi.org/10.1021/ba-19791072 0177.ch002.
- 1073 (98) Akerlof, G. Activity Coefficients of Sodium, Potassium and Lithium Chlorides and
 1074 Hydrochloric Acid at Infinite Dilution in Water-Methyl Alcohol Mixtures. *J. Am. Chem.*1075 Soc. 1930, 52, 2353–2368.
- 1076 (99) Yan, W.; Xu, Y.; Han, S. Activity Coefficients of Sodium Chloride in Methanol-Water
 1077 Mixed Solvents at 298.15 K. *Acta Chim. Sin.* 1994, 52 (10), 937-946 (in Chinese).
- 1078 (100) Deyhimi, F.; Abedi, M. NaCl + CH 3OH + H 2O Mixture: Investigation Using the Pitzer
 1079 and the Modified Pitzer Approaches to Describe the Binary and Ternary Ion-
- 1080 Nonelectrolyte Interactions. J. Chem. Eng. Data **2012**, 57 (2), 324–329.
- 1081 https://doi.org/10.1021/je201084a.
- (101) Kozłowski, Z.; Bald, A.; Gregorowicz, J. Thermodynamic Studies of NaCl Solutions in
 Water + Methanol Mixtures by Means of a Galvanic Cell Containing a Glass Sodium
 Electrode. J. Electroanal. Chem. 1990, 288 (1–2), 75–82. https://doi.org/10.1016/00220728(90)80026-3.
- 1086 (102) Jödecke, M.; Kamps, Á. P. S.; Maurer, G. Experimental Investigation of the Influence of
 1087 NaCl on the Vapor-Liquid Equilibrium of CH3OH + H2O. *J. Chem. Eng. Data* 2005, *50*1088 (1), 138–141. https://doi.org/10.1021/je049783k.
- 1089 (103) Sun, R. Calculation of Salting Effect on Vapor-Liquid Equilibrium in Water-
- 1090 Nonelectrolyte Systems. *Chem. Eng.* **1995**, *23* (1), 13–17, 30 (in Chinese).
- 1091 (104) Abedi, M.; Shendi, M. A.; Karimzadeh, Z.; Salamat-Ahangari, R.; Deyhimi, F. The

- Ternary and Quaternary Electrolyte Systems: Activity Coefficient of NaCl Measured and
 Modeled for NaCl + C2H5OH + H2O and NaCl + KCl + C2H5OH + H2O Systems. *Fluid Phase Equilib.* 2016, 423, 138–145. https://doi.org/10.1016/j.fluid.2016.04.023.
 (105) Chen, W.; Zhang, W. Vapor-Liquid Equilibria for Alcohol-Water-KI/NaAc Systems. J.
- *Chem. Eng. Chinese Univ.* 2003, *17* (2), 123-127 (in Chinese).