

# <sup>1</sup> Battflow: an automated workflow for predicting key properties of battery electrolytes

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## Software

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## <sup>6</sup> Summary

<sup>7</sup> Battflow provides an automated workflow that integrates a suite of Python packages,  
<sup>8</sup> GROMACS ([Abraham et al., 2025](#)), and ORCA ([Neese, 2012](#)) to predict properties of battery  
<sup>9</sup> electrolytes. It aims to simplify user interactions when generating large MD/DFT datasets,  
<sup>10</sup> thereby aiding the high-throughput screening and discovery of new electrolytes. The workflow  
<sup>11</sup> is linked to a collection within a MongoDB cluster, in which missing calculated properties  
<sup>12</sup> are identified and flagged for further calculations. The only information needed to start the  
<sup>13</sup> simulations is the SMILES string and molar concentration of each electrolyte component. A  
<sup>14</sup> document for each electrolyte composition is then created, and diffusion coefficients and  
<sup>15</sup> solvation structure statistics are computed from molecular dynamics simulations, while binding  
<sup>16</sup> energies for solvation clusters and HOMO–LUMO energies of each individual component are  
<sup>17</sup> updated.

## <sup>18</sup> Statement of need

<sup>19</sup> Lithium metal batteries (LMBs) are regarded as a promising solution for meeting market  
<sup>20</sup> demand for energy storage systems with high specific capacity. Recently, several battery  
<sup>21</sup> technologies incorporating lithium metal anodes have attracted increasing attention, including  
<sup>22</sup> lithium–sulfur (Li–S) batteries, lithium–oxygen (Li–O<sub>2</sub>), and lithium–carbon dioxide (Li–CO<sub>2</sub>)  
<sup>23</sup> batteries. However, the implementation of lithium metal anodes is currently hindered by  
<sup>24</sup> poor cycle life and uncontrollable side reactions between Li metal and liquid electrolytes.  
<sup>25</sup> Liquid electrolyte engineering, which involves mixing different molecules to create electrolytes  
<sup>26</sup> with specific properties, is ultimately the most cost-effective approach for making LMBs  
<sup>27</sup> viable. However, there are myriad possible electrolyte formulations due to the large number  
<sup>28</sup> of commercially available molecules, recently synthesised electrolyte-specific compounds, and  
<sup>29</sup> various strategies for fine-tuning electrolyte components. Consequently, there is an increased  
<sup>30</sup> need for theory-guided rational design of new electrolyte formulations for LMBs, aiming to  
<sup>31</sup> reduce research costs and avoid “trial-and-error” approaches.

<sup>32</sup> There is an increasing need for standardised computational data to guide experimental studies  
<sup>33</sup> on battery electrolytes, particularly in light of the rapid growth of the literature in this field,  
<sup>34</sup> with approximately 200 papers containing the keywords “battery electrolyte” published per  
<sup>35</sup> week in 2023 alone. A combination of molecular dynamics (MD) and Density Functional  
<sup>36</sup> Theory (DFT) simulations for the estimation of transport and electronic properties of the bulk  
<sup>37</sup> electrolyte and its individual components offers a good balance between accuracy and cost  
<sup>38</sup> efficiency for properties prediction. However, setting up force fields and simulation settings for  
<sup>39</sup> molecular dynamics, followed by DFT calculations of relevant Li solvation clusters for several  
<sup>40</sup> electrolyte compositions, can be a daunting task and error-prone task, even for experienced  
<sup>41</sup> researchers.

42 Battflow is intended for both theoreticians and experimentalists and can provide out-of-the-box  
43 default settings to run the automated workflow with basic configurations. At present, Battflow  
44 extends beyond Li-metal and Li-ion batteries and can be adapted to other alkali metal battery  
45 systems, including Na-ion, K-ion, and Zn-based batteries.

## 46 Usage and availability

47 Battflow inputs consist of documents stored within a MongoDB collection. The information  
48 required to connect to MongoDB, either through localhost or a remote instance, is provided  
49 in config.yaml. A .json example input file, which should be uploaded to MongoDB as a  
50 document, is provided. Battflow reads the smiles and concentrations fields within each  
51 document to build the molecular structure of each component and to create the electrolyte box,  
52 respectively. The output is reported in the simulation\_data field; if any calculated property  
53 is flagged as absent, the workflow is triggered to start.

54 The molecular dynamics workflow consists of: (1) setup of GAFF2 force fields for the molecular  
55 components, generated using ACPYPE ([Sousa da Silva & Vranken, 2012](#)); (2) creation of  
56 the electrolyte box; (3) execution of minimisation, equilibration, and production MD runs  
57 using GROMACS ([Abraham et al., 2025](#)); and (4) uploading diffusion properties and solvation  
58 structure statistics to the MongoDB document. The DFT simulations in ORCA ([Neese, 2012](#))  
59 follow the molecular dynamics runs, with calculations performed for the three most prevalent  
60 solvation clusters. Binding energies are computed for every solvation cluster and HOMO-LUMO  
61 energies are calculated for the entire cluster and separated components, as well. After the  
62 analysis, results are uploaded back to MongoDB.

63 Battflow is available for Linux operating systems and can be downloaded from GitHub  
64 (<https://github.com/neubifx/Battflow/tree/main>) under the GPL-3.0 licence. Additional  
65 documentation is available on the repository page and is continuously updated.

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