Testing the Limits of Urban's Method for Voltage Calculation

Motivation

Lithium-ion batteries (LIB) show much promise for energy storage mechanisms but Urban's method to calculate voltage applies to the following are far from being perfected. Sensitivity to ambient oxygen and moisture, low energy chemical reaction with the equation written below: density, dendrite growths, and short battery life are all problems that plague LIB on the market today. To solve these issues, the Mo group uses atomic-scale modeling of LIB materials to gain insight on reaction pathways of both existing and theoretical E(Lix1MO2) - E(Lix2MO2) - (x1 - x2)E(Li)materials. One aspect of this research is on the voltage of cathode materials, which is (x1 - x2)the subject of this project. In 2016, Urban et al published a paper studying the voltages for lithium metal oxides of the form LiMO2, and their thermodynamic Literature voltage values were provided by the Materials Project, stabilities against various electrolytes was further studied by Nolan et al (2019). This which has already done full MDS with the Vienna Ab-initio project examined whether this formula could be extended to polyatomic oxyanions Simulation Package (VASP) to produce highly accurate results. Total and found that formula-calculated results were highly precise but reliably 1.0V below energy values were also given by the Project, and these were the the theoretical value for that average. On its own Urban's method was not accurate values put into Urban's formula. Percent error was then evaluated but a slightly modified version of it would work phenomenally. Usage of this much to judge the accuracy of the simple calculation method. Four simpler formula will allow future researchers to gain a reasonably accurate estimate different types of polyatomic oxides were examined: phosphates, of voltage of a theoretical material without needing to do a full Molecular Dynamics sulfates, silicates, and carbonates. Many transition metals were Simulation (MDS), which saves time, power, computer memory, and other relevant involved, the most abundant sample sizes being Fe, Mn, Co, and Cr. resources.



Citations

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Abstract

$$\operatorname{Li}_{x_1}MO_2 \rightarrow \operatorname{Li}_{x_2}MO_2 + (x1-x2)\operatorname{Li}$$
 for any $x1 > x2$



Zhou, F., et al. "First-Principles Prediction of Redox Potentials in Transition-Metal Compounds with \$\mathrm{LDA}+U\$." Physical Review B, vol. 70, no. 23, Dec. 2004, p. 235121. APS, https://doi.org/10.1103/PhysRevB.70.235121.

Transition Metals

The data on transition metals was much less clear-cut than that of the oxide categories. Much of this was because there were many more metals involved in the overall data pool than there were oxide types, meaning that the sample size for each metal was much smaller. The metals with the greatest sample sizes were Fe, Co, and Mn, closely followed by Cr. Other metals included in this project were V, Ti, Ni, and Cu; the latter two are popular for use in Li batteries. More work will be needed to study these categories in depth because I did not have enough time to do this myself.

Metal	Average	Μ
Со	25.72	7
Cr	21.52	-
Cu	11.21	0
Fe	19.80	5
Mn	25.95	-
Ni	2.494	0
Si	0.2761	0
Ti	4.655	3
V	9.048	Ę
Zn	21.12	Ĩ

Table 1. Statistics for each transition metal.

Figure 3. Table 1 shown in a bar graph. Medians consistently lower than the average show that all sets of data have a positive skew of error, and that most metals except Mn and Cr can be trusted. Judgement should be withheld for Zn and Si because their sample sizes are too small to establish a pattern.

Conclusions:

This study showed that Urban's method was generally accurate but not enough to be completely relied on. If one is evaluating the voltage for a material not within the Materials Project database, they would be better off not using Urban's method at all and doing the full MDS, even though such a computation is much more resource-intense. It could still be useful to relax the crystals and use Urban's method to gain a general ballpark, but the outcome of the equation should by no means be prioritized over the VASP-computed values. Sources of systematic error included carbonates, which contained the highest amount of egregiously incorrect values, but beyond that there are no clear-cut sources of either systematic nor random error. The median percent error for all tested groups was much lower than the averages, indicating that the calculated values were generally close to the theoretical voltage. The average error was not a good metric because one or two outliers could position the average at double the median and would therefore make it not representative of the data set it belonged to. More research could certainly be done to expand the data, especially to include more robust counts of the transition metals discussed above.

Lessons Learned

There are several lessons I will take away from this REU. The first is that materials research is a prime example of the idiom 'easier said than done.' With a system as delicate as a LIB, there are many checkboxes a material (or combination of materials) must meet to be considered a viable candidate for testing in a physical laboratory environment. Finding a new material to fit the job is like playing whack-a-mole. For example, LGPS has high efficiency and energy density but is highly reactive to ambient moisture; LiFePO₄ (LFP) is stable and lightweight but has a lower energy density; and so on. This has been very eye-opening in terms of the open-ended problem-solving process of research, and that there isn't necessarily a 'right answer' like there might be to the worksheets and homework one will see in classroom settings. Additionally, I have gained experience in tools that are popular in chemical and material research. Pymatgen and VASP are key tools for computational modelling of all sorts of substances and not just LIB, which makes them transferrable to other research positions that I might hold in the future.

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