

A New Paradigm for Anionic Water Pollutant Adsorption: High-Entropy LDHs

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Motivation

- Nuclear fission energy: 10% of the entire energy generated, 30% for Korea
- Radioactive anions released

Isotope	Half-Life
Cs-137	~ 30 years
Sr-90	~ 28.8 years
I-129	~ 15.7 million years
I-131	~ 8 days
Tritium	~ 12.3 years
Pu-239	~ 24,100 years

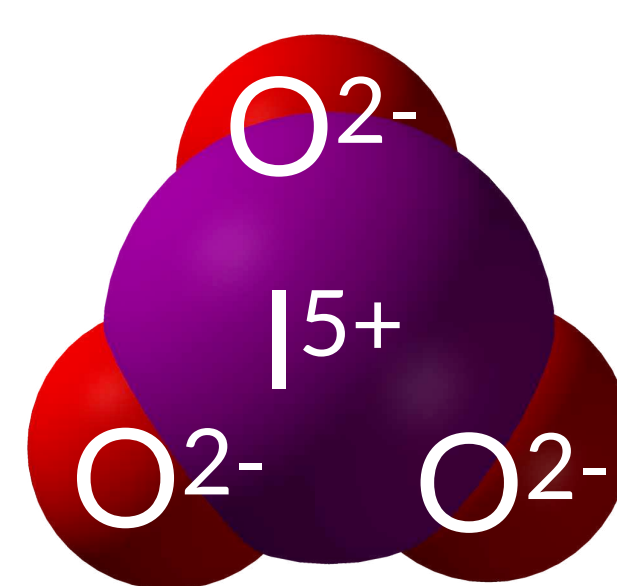


Figure 1. Iodate

Table 1. 7 long-lived radioactive isotopes of nuclear fission

- I-129 is present as IO_3^- in coolant.
- Iodine has outstanding half-life.
- Artificial adsorption is required.

Table 2. Previous methods and their problems

Previous Methods	Problems
Activated carbon	IO_3^- adsorption rate of 29.9% Need to acidify processing water to adsorb
Ag-zeolite	Extremely expensive IO_3^- adsorption rate of 3.6%

Layered Double Hydroxide, LDH

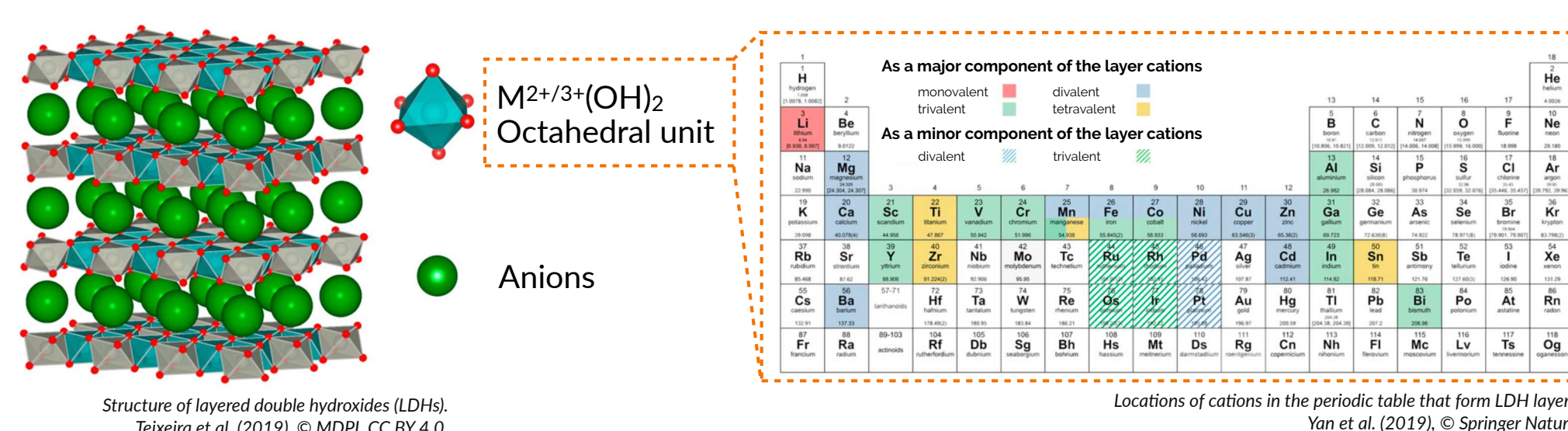
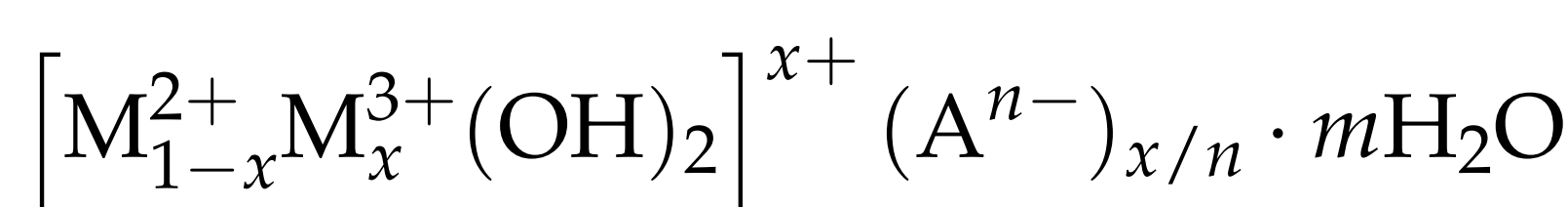


Figure 2. Layered Double Hydroxide, LDH



- Flexibly designable and cheap
- Anions can be captured through surface adsorption and anionic exchange
- Insufficient maximum adsorption rate of under 90%

Goal

Increase the IO_3^- adsorption rate!

Approach

High-Entropy LDH

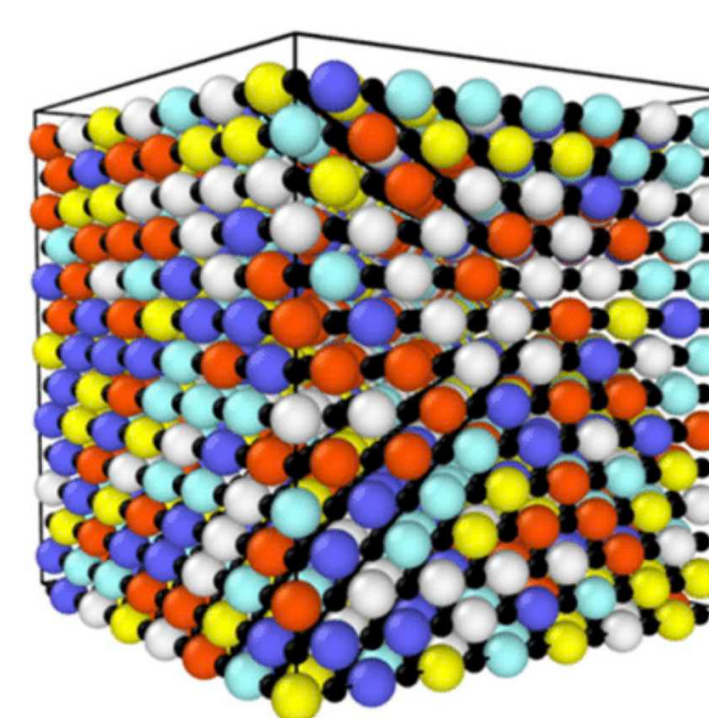


Figure 3. High-entropy material

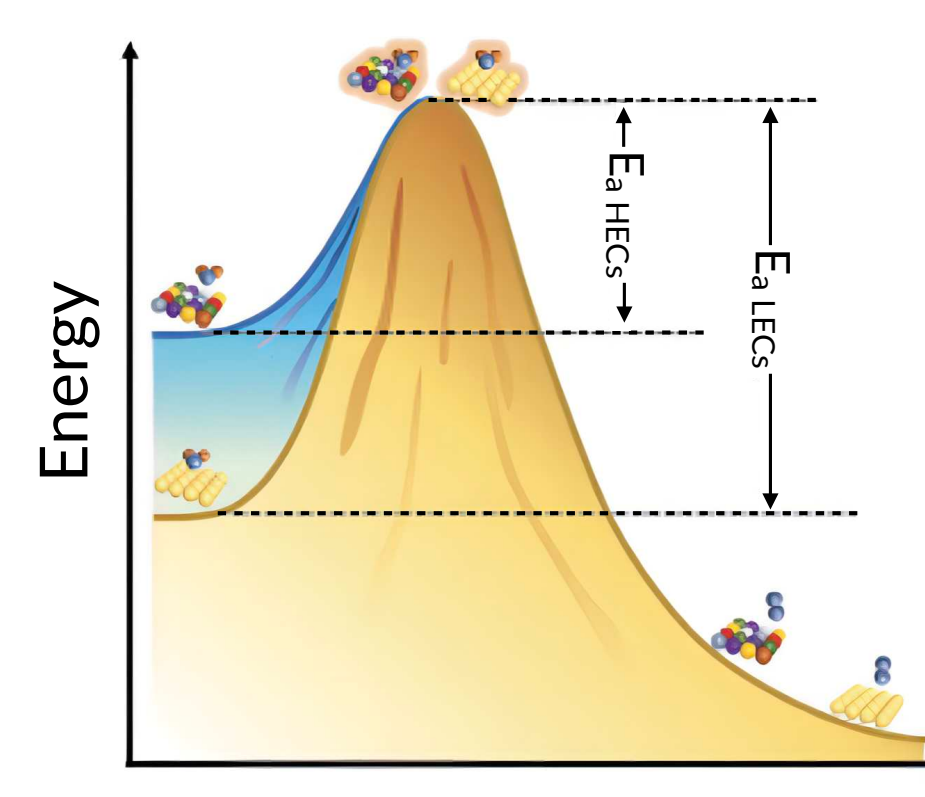


Figure 4. Reduction of activation energy

- 5+ elements mixed at 5-35 at.%
- Stable and multifunctional
- Reduces activation energy to adsorb
- Creates abundant active sites

Experimental Process

- Find optimal composition of high-entropy LDH via machine learning
- Synthesize high-entropy LDH and measure its IO_3^- adsorption rate
- Verify the validity of the approach

Experiment 1: Predicting LDH Compositions for Iodate Adsorption via ML

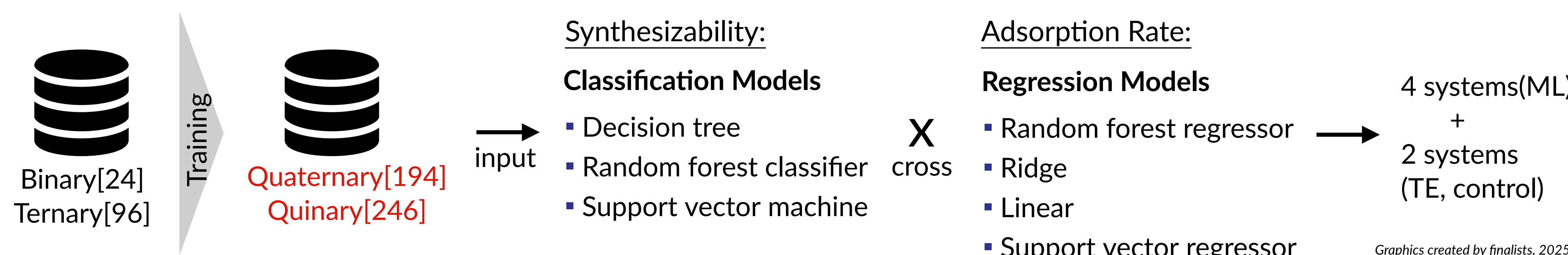


Figure 5. Machine learning process for LDH composition determination

Table 3. Selected LDH compositions

Sample No.	Composition	Selection Method
#4.1	$(\text{NiCu})_3(\text{CrAl})_1$	Machine learning
#4.2	$(\text{Cu})_3(\text{CrFeAl})_1$	Machine learning
#4.3	$(\text{NiCu})_3(\text{FeAl})_1$	Trial and Error
#5.1	$(\text{CoNiCu})_3(\text{CrAl})_1$	Machine learning
#5.2	$(\text{NiCu})_3(\text{CrFeAl})_1$	Machine learning
#5.3	$(\text{NiZnCu})_3(\text{FeAl})_1$	Trial and Error

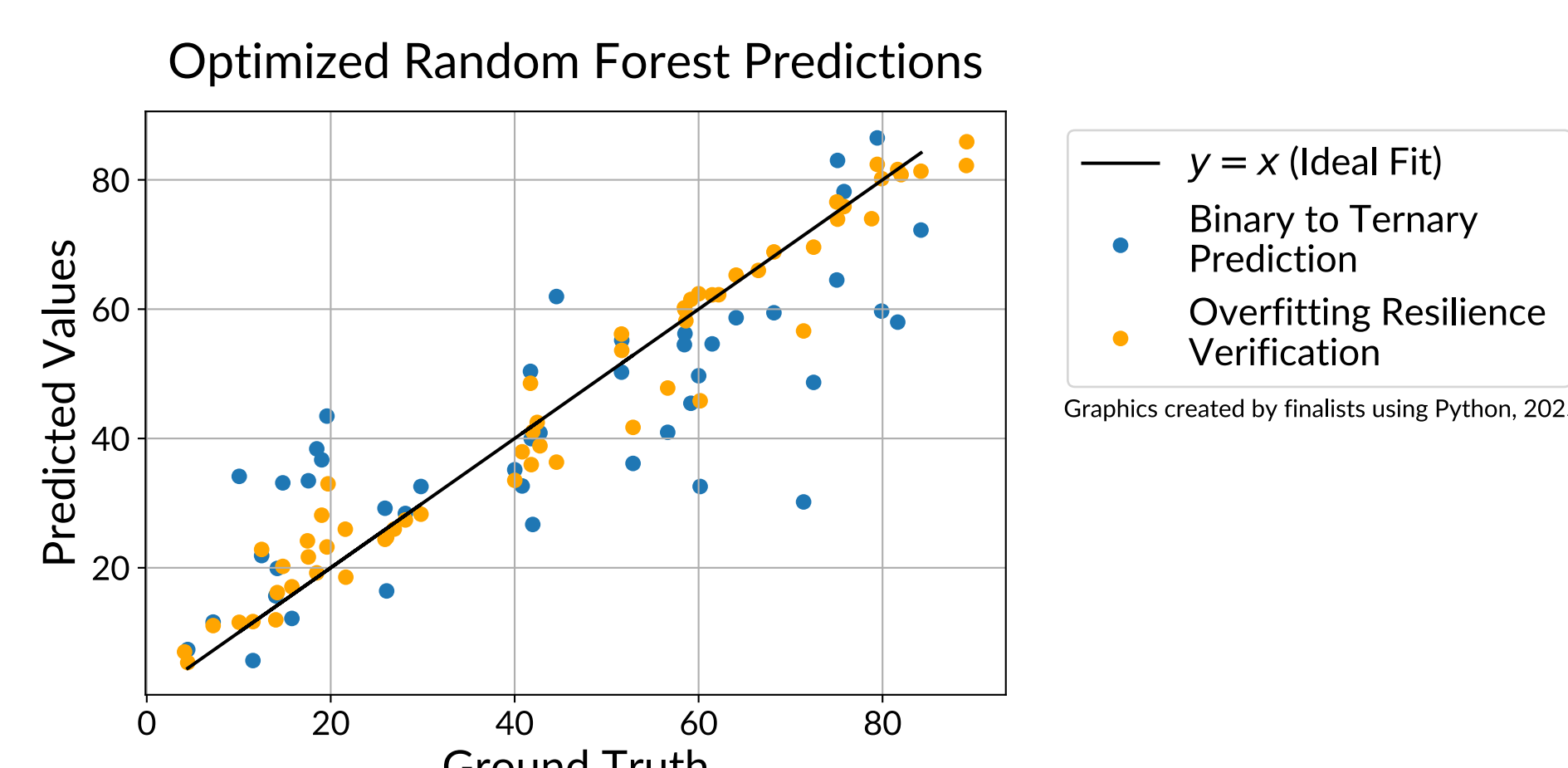


Figure 6. Machine learning's validity

- Blue dots: validity of machine learning
- Yellow dots: No signs of overfitting observed

Experiment 2: Synthesis and Verification of LDHs

1. LDH Synthesis (Co-precipitation)

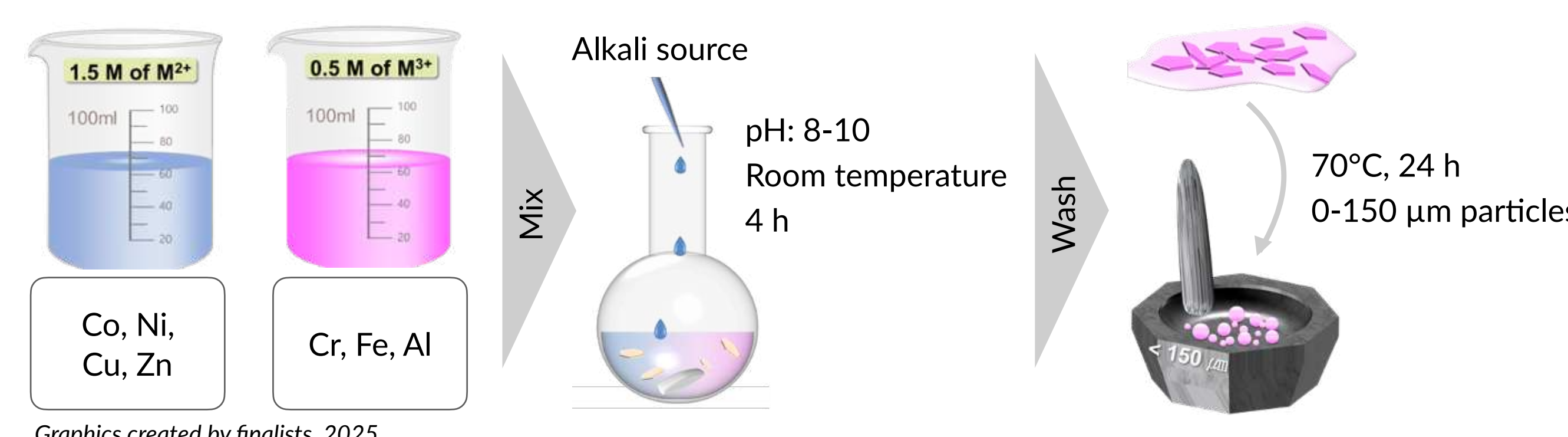


Figure 7. LDH synthesis using co-precipitation method

2. LDH Synthesis Verification

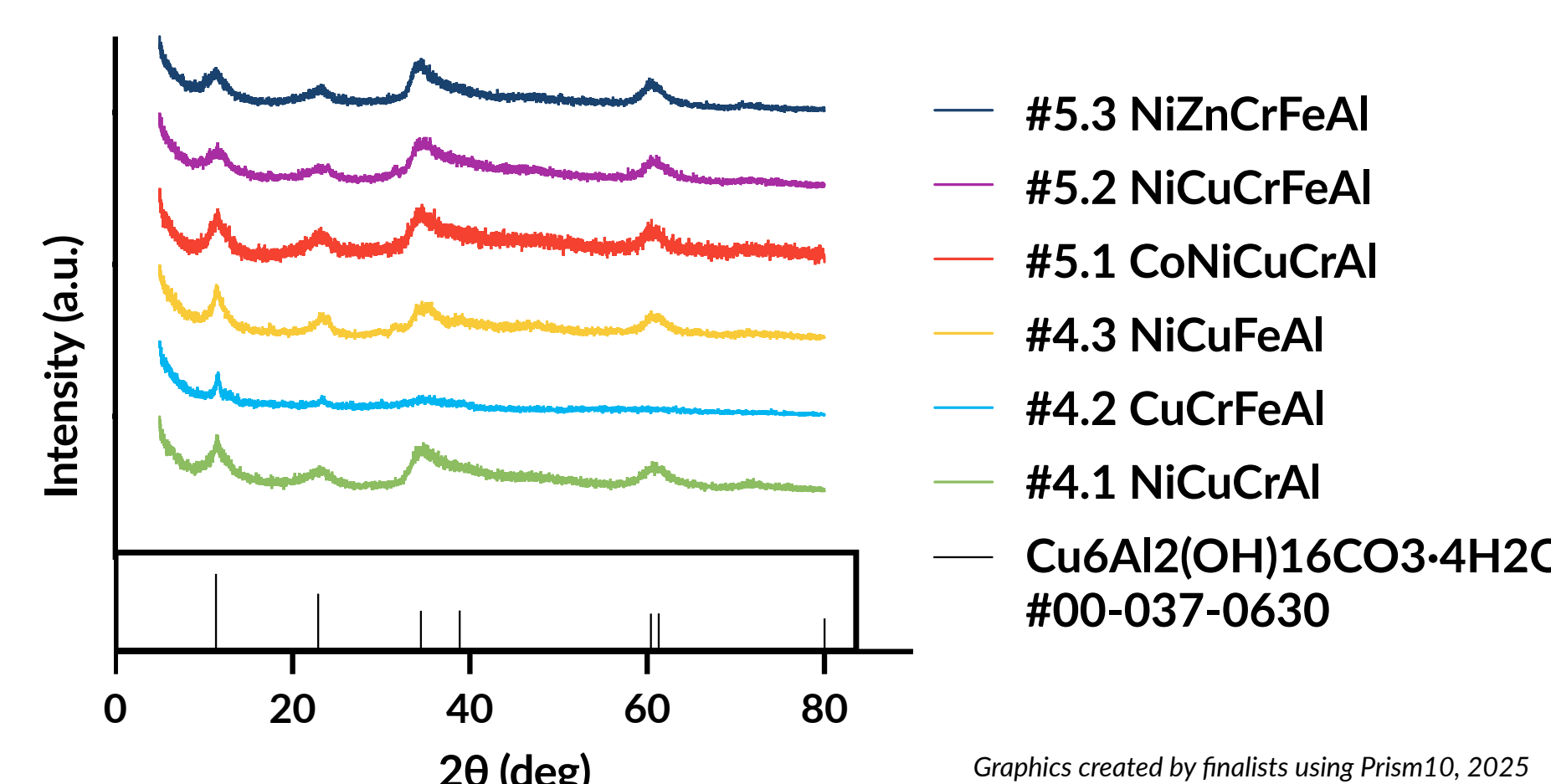


Figure 8. Comparison of powder XRD patterns

- Figure 8 illustrates the successful synthesis of LDH by comparing the peaks.
- Figure 9 shows that the synthesis was done with designed ratio: $[M^{2+}] : [M^{3+}] = 3 : 1$ and equimolar within metals of same ionic charge.

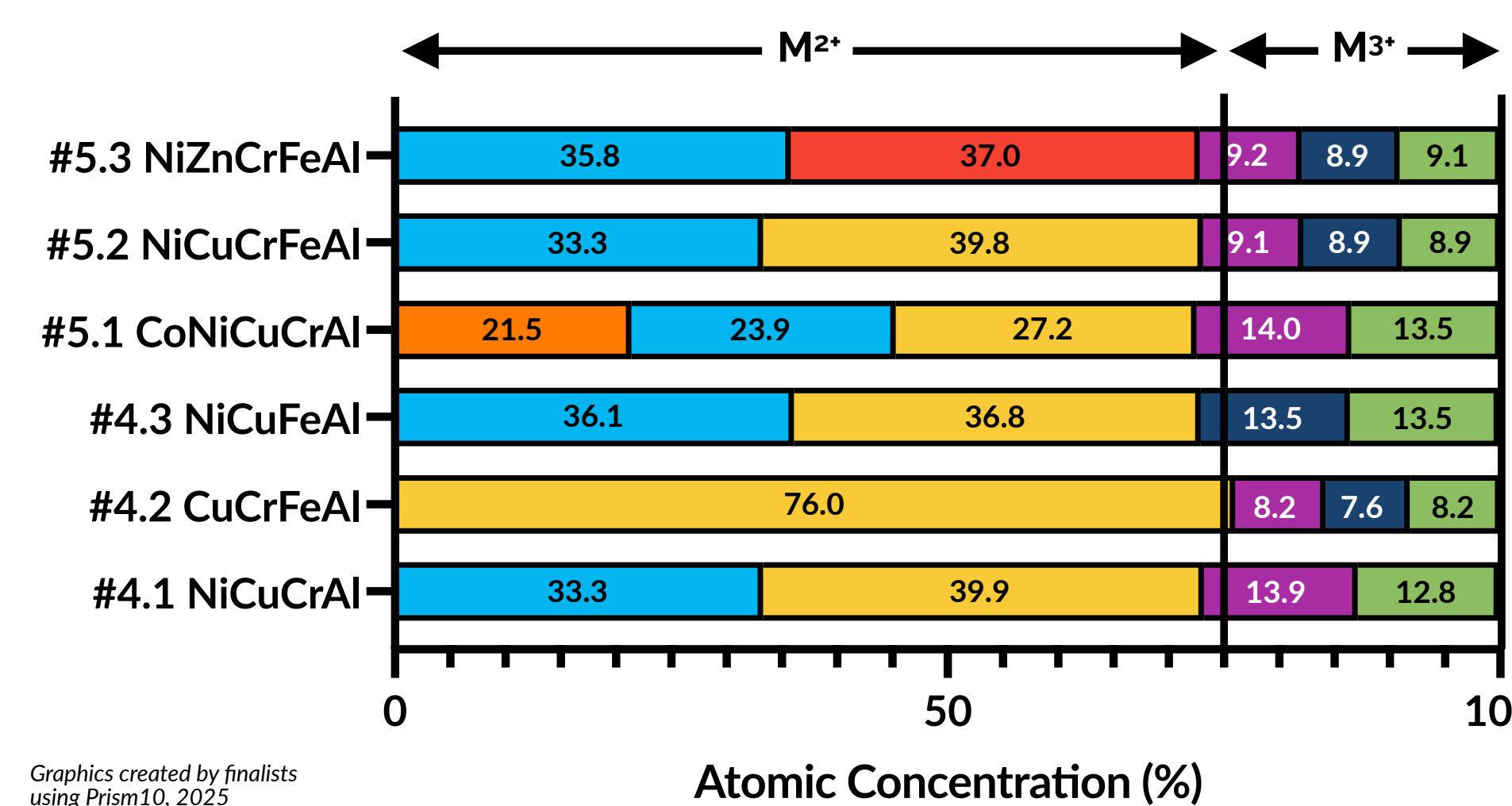


Figure 9. Ratio of elements in compositions

Experiment 3: Iodate Adsorption Performance Measurement

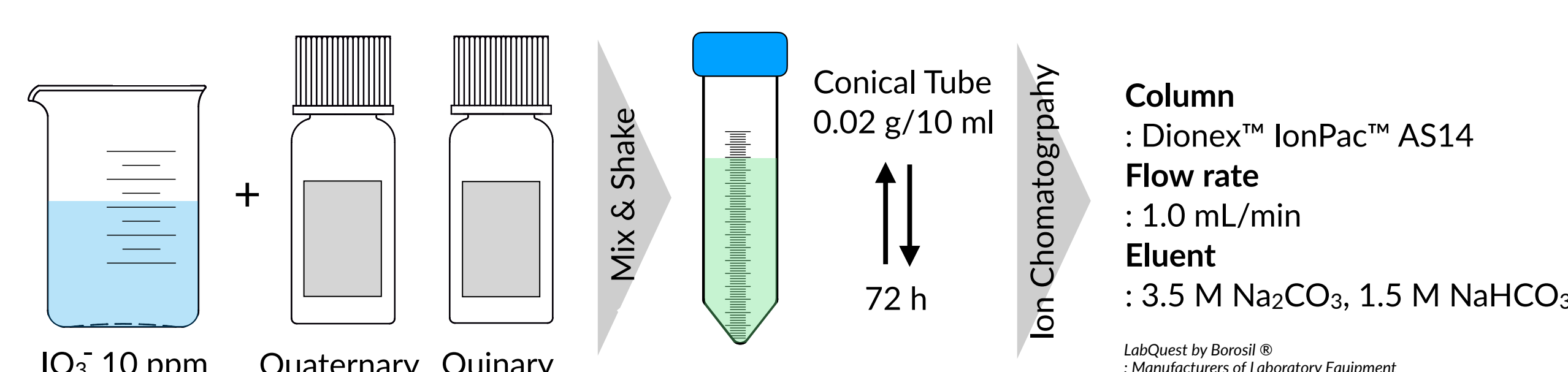


Figure 10. IO_3^- adsorption experiment process

- Adsorption rate is convergent after adsorbing for 48 hours.
- 72 hours of adsorption is enough.

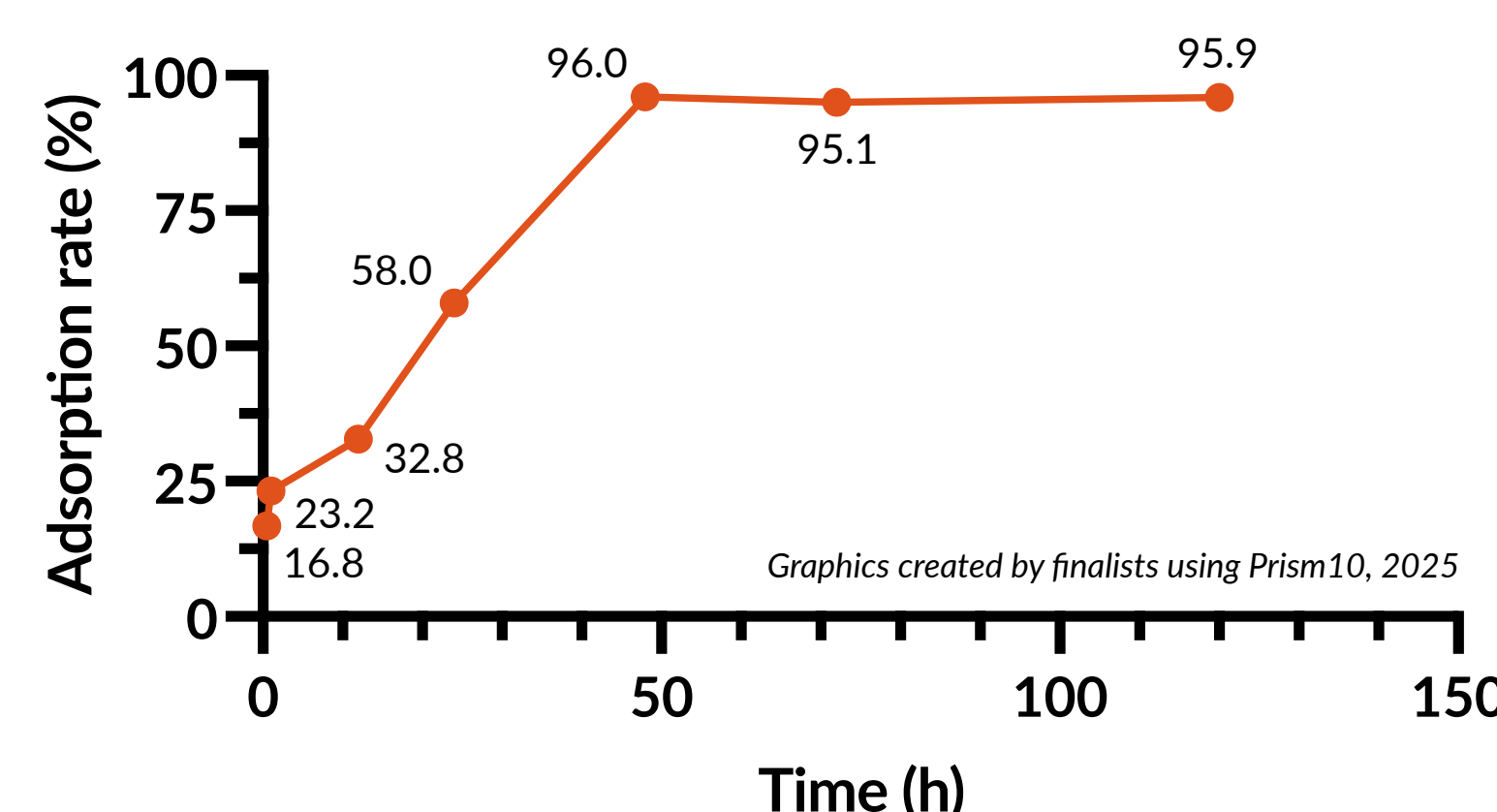


Figure 11. IO_3^- adsorption kinetics of #4.2 CuCrFeAl under a same condition

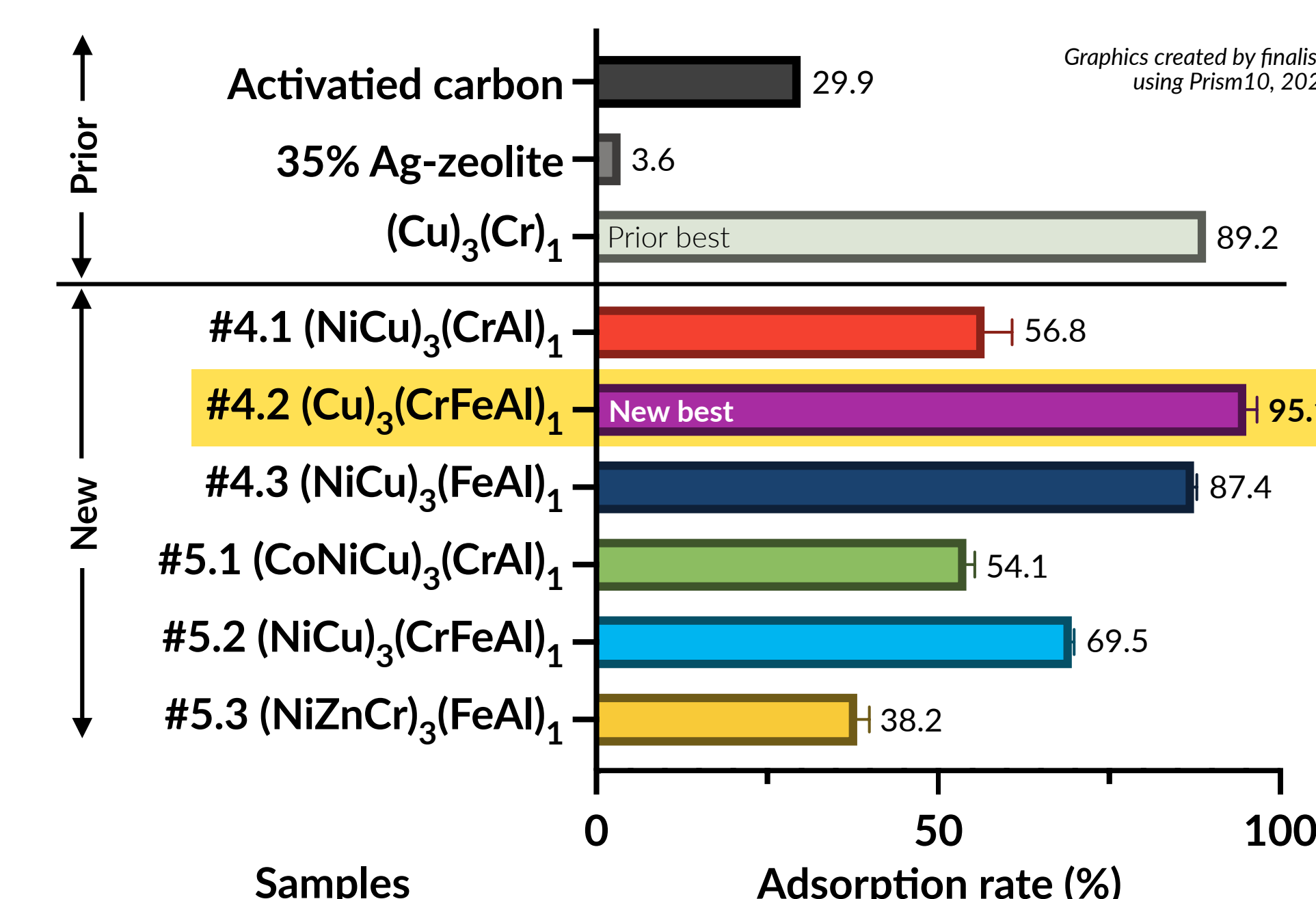


Figure 12. Comparison of IO_3^- adsorption rates with previous methods

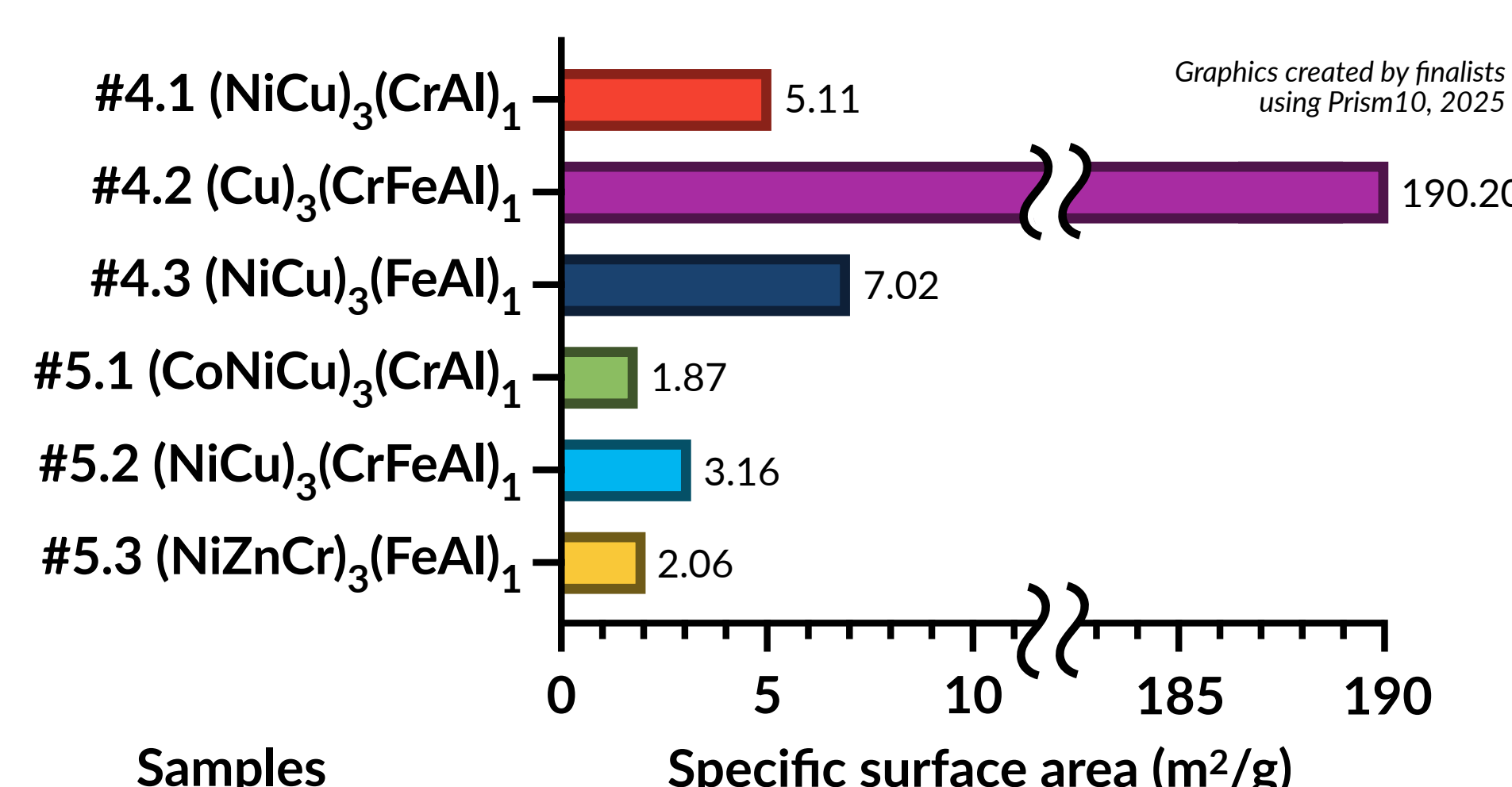


Figure 13. Specific surface area of LDH

Discussion

► Analysis of the results

- The specific surface area and adsorption rate show a similar trend. (Figure 13 and Figure 12)
- : Low crystalline materials have more surface defects, resulting in a higher surface area.
- Powder XRD indicates that sample #4.2 exhibits low crystallinity. (Figure 8)
- : Higher surface area leads to more adsorption.

$$\therefore \frac{1}{\text{Crystallinity}} \propto \frac{\text{Surface area}}{\text{Mass}} \propto \text{Adsorption rate}$$

► So, how can we maximize the adsorption rate?

- Expand the idea to high-entropy effect.
- Balance the composition.
- Find optimized number of elements.

The configurational entropy is:

$$\Delta S_{\text{quaternary}} = -R \sum_{i=1}^n x_i \ln x_i \geq 1.39R,$$

where R is the gas constant.

Even in quaternary LDHs, the entropy increase is sufficiently high to exhibit high-entropy effects.

$\therefore$ Approach is proven valid.

Applications

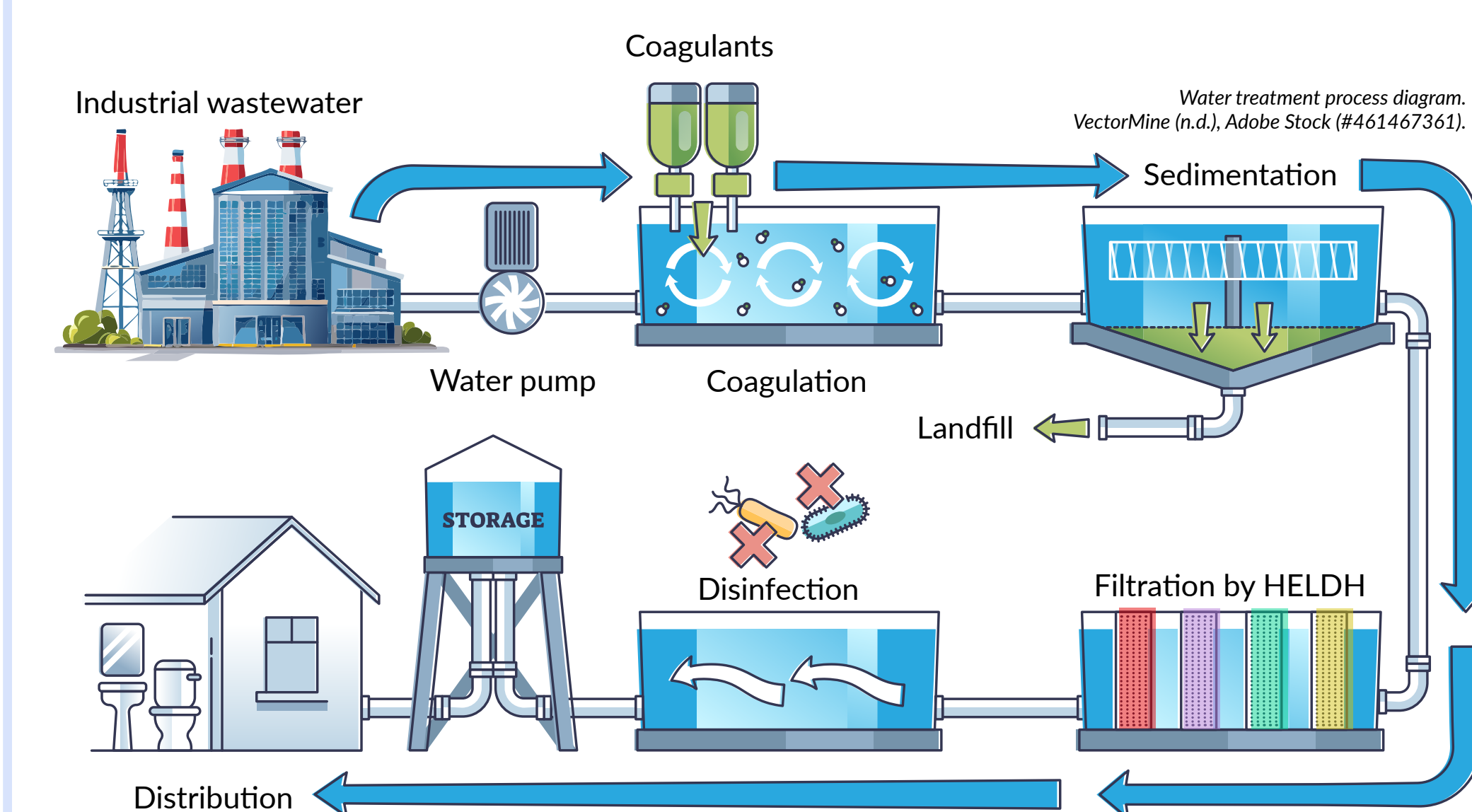


Figure 14. LDH filtration of industrial wastewater

- Textile, paper, chemical manufacturing industries release wastewater that contains IO_3^- .
- LDH fixation into filter helps adsorb flowing anionic pollutants.

Conclusion

- Expanding an idea to high-entropy LDH led to a remarkable improvement in performance.
- $(\text{Cu})_3(\text{CrFeAl})_1$ was discovered following this expansion, increasing the previous maximum from 89.2% to 95.1%.
- It will pave the way to adsorb a broader range of anionic pollutants, ultimately contributing to environmental preservation.

References

- Deng, C. et al. (2024). High entropy materials for catalysis: A critical review of fundamental concepts and applications. Nano Energy, 120, 109153.
- Saju, D. et al. (2020). Layered Double Hydroxides: Fundamentals to Applications. Layered Double Hydroxide Polymer Nanocomposites, 1–76.
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