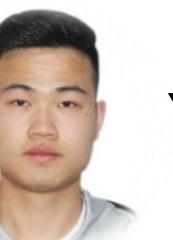


MolKGNN: Interpretable Chirality-Aware Graph Neural Network for Drug Discovery



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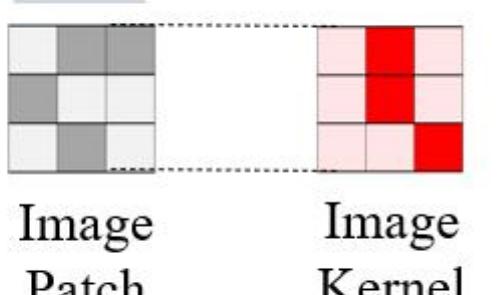
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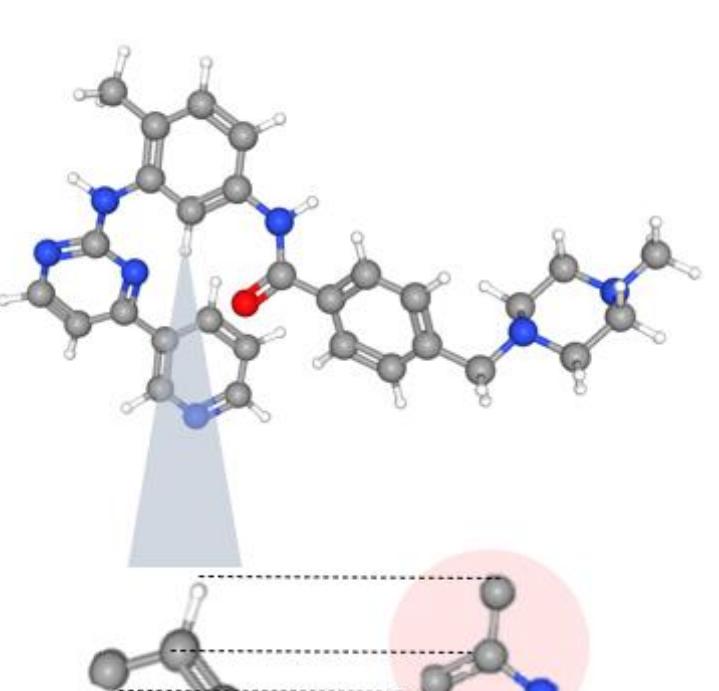
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ABSTRACT



(A)



(B)

Fig. 1 MolKGNN extends convolution specifically for drug discovery

METHOD

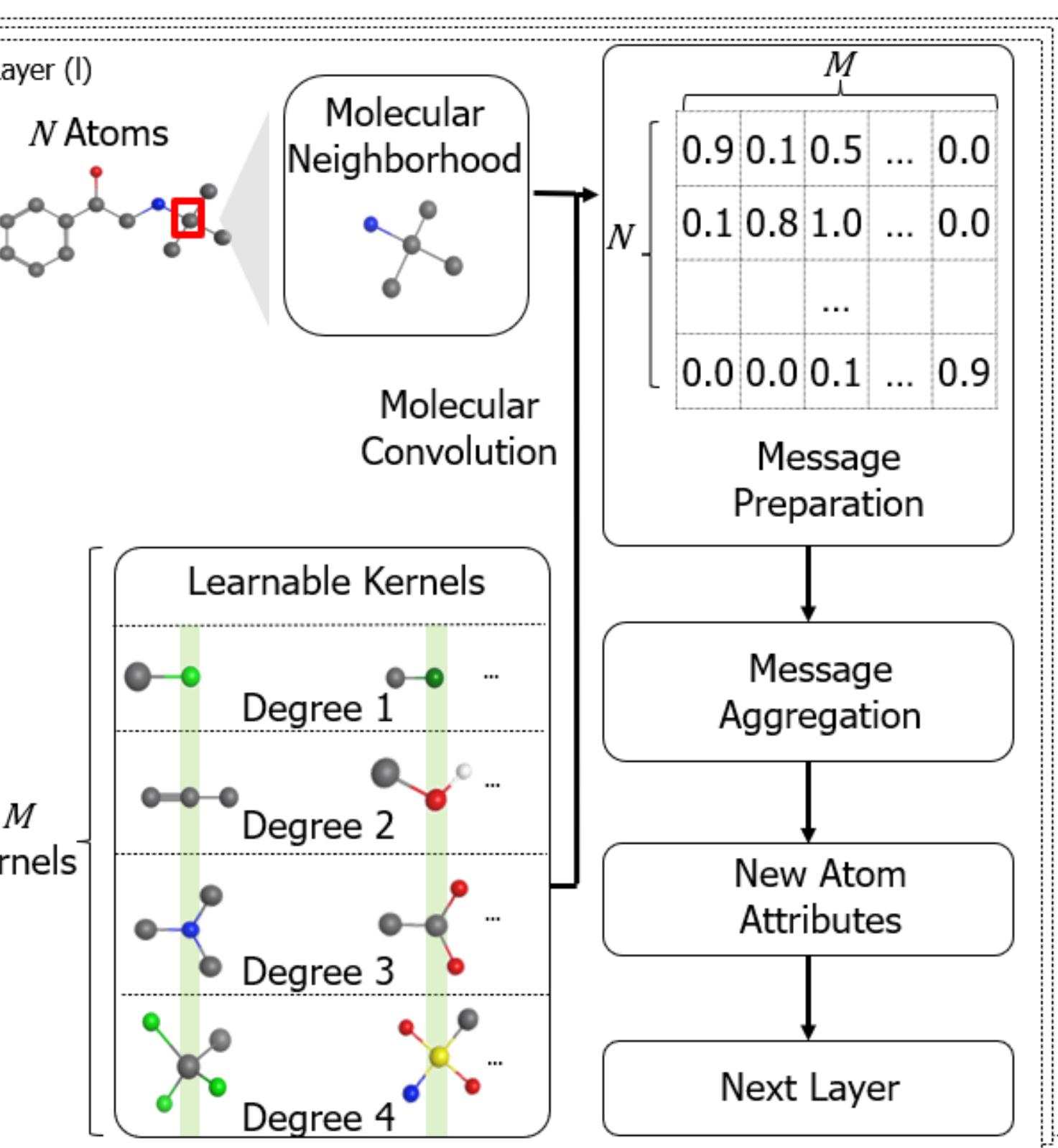


Fig. 4 Method overview

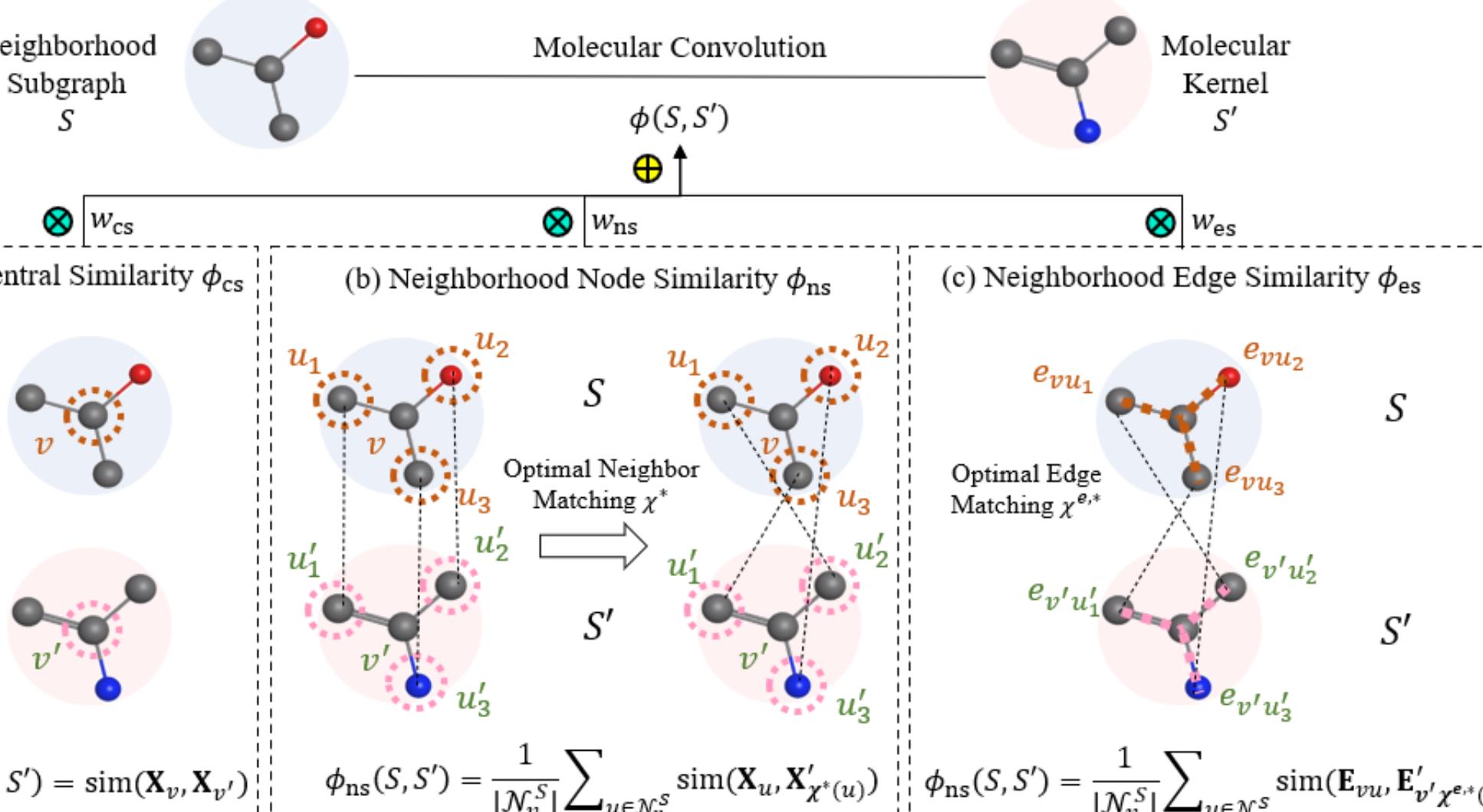


Fig. 5 Similarity score calculation

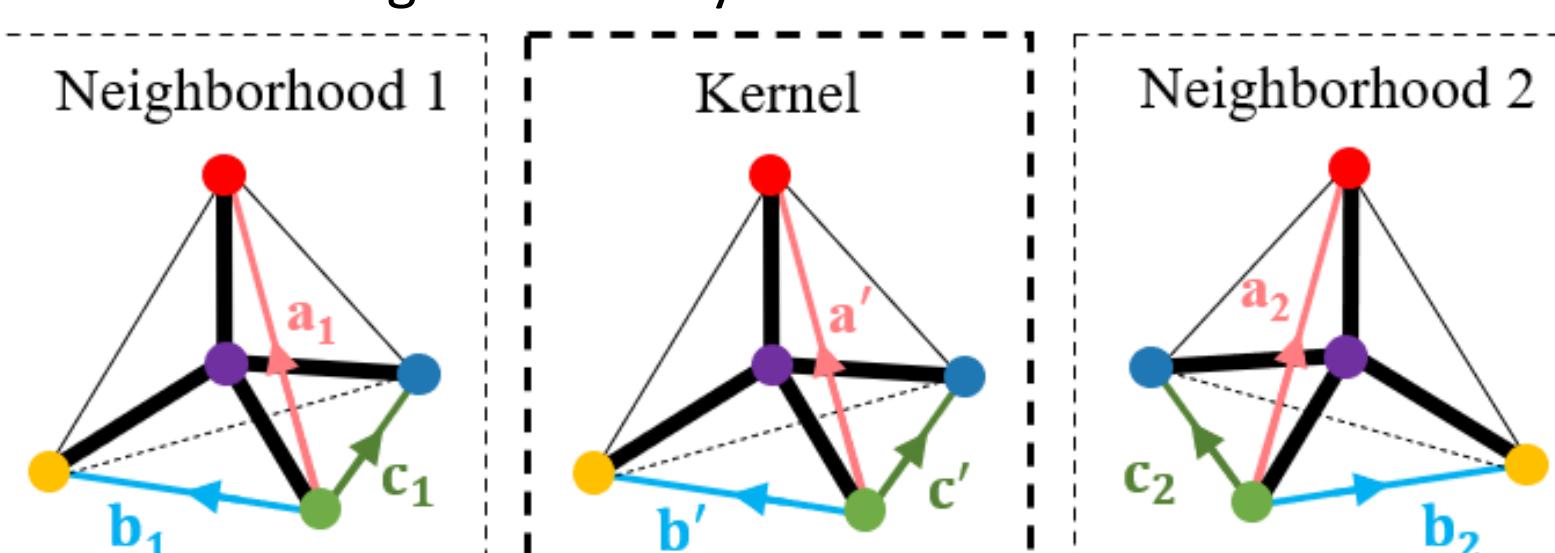
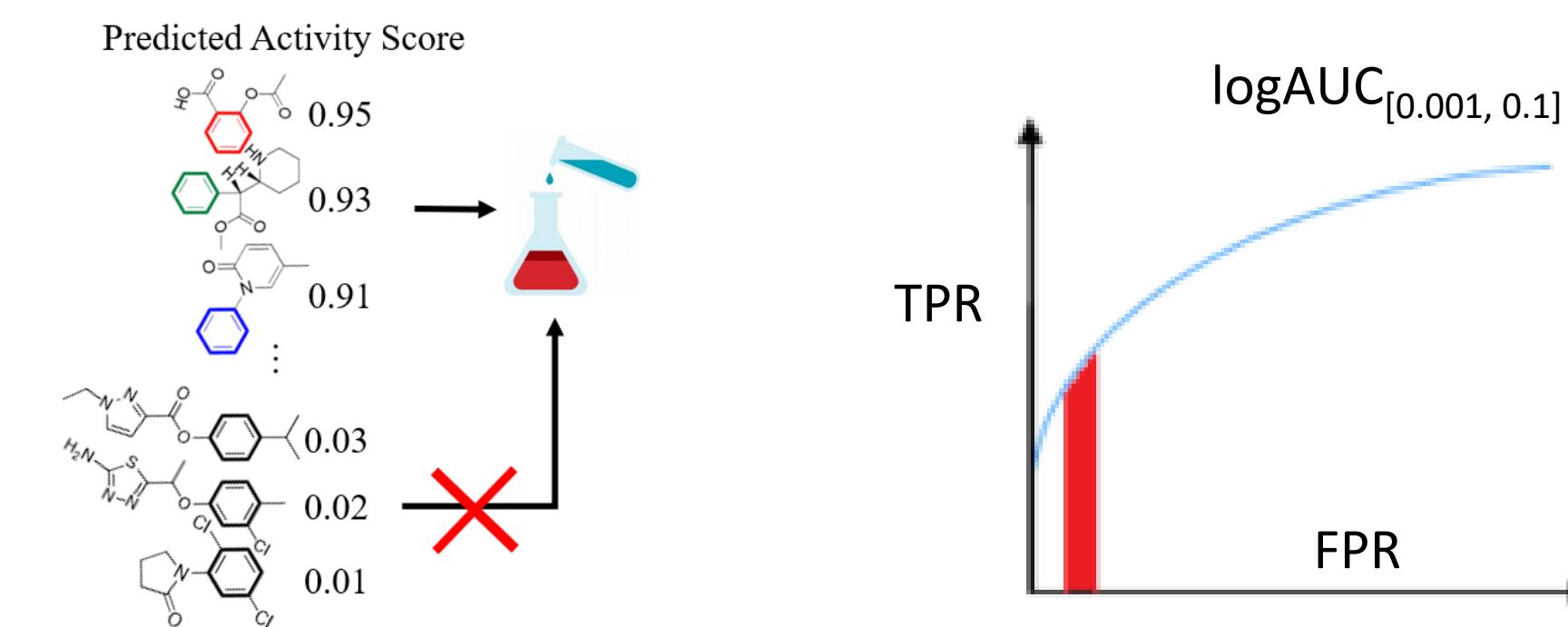

Sgn($\mathbf{a}_1 \times \mathbf{b}_1 \cdot \mathbf{c}_1$) = Sgn($\mathbf{a}' \times \mathbf{b}' \cdot \mathbf{c}'$) ≠ Sgn($\mathbf{a}_2 \times \mathbf{b}_2 \cdot \mathbf{c}_2$)

Fig. 6 Chirality Calculation

DOMAIN-RELEVANT METRIC


Fig. 7 Only top-ranked predictions are being experimentally validated due to the cost. $\log\text{AUC}_{[0.001, 0.1]}$ is used to bias toward the performance on these top-ranked predictions

REPRESENTATIVE RESULT

Tab. 1 $\log\text{AUC}_{[0.001, 0.1]}$ Performance (↑)

Dataset ¹	MolKGNN	KerGNN ¹	ChiRo ²	SphereNet ³
435008	0.25±0.01	0.15±0.01	0.17±0.02	0.22±0.02
1798	0.17±0.03	0.08±0.04	0.17±0.04	0.20±0.04
435034	0.23±0.02	0.18±0.05	0.21±0.02	0.23±0.03
1843	0.36±0.03	0.29±0.03	0.33±0.01	0.26±0.05
2258	0.30±0.03	0.19±0.02	0.25±0.01	0.38±0.04
463087	0.39±0.06	0.15±0.01	0.26±0.02	0.40±0.01
488997	0.30±0.03	0.08±0.02	0.19±0.03	0.31±0.03
2689	0.42±0.02	0.26±0.02	0.35±0.05	0.40±0.02
485290	0.50±0.01	0.22±0.03	0.29±0.07	0.45±0.04

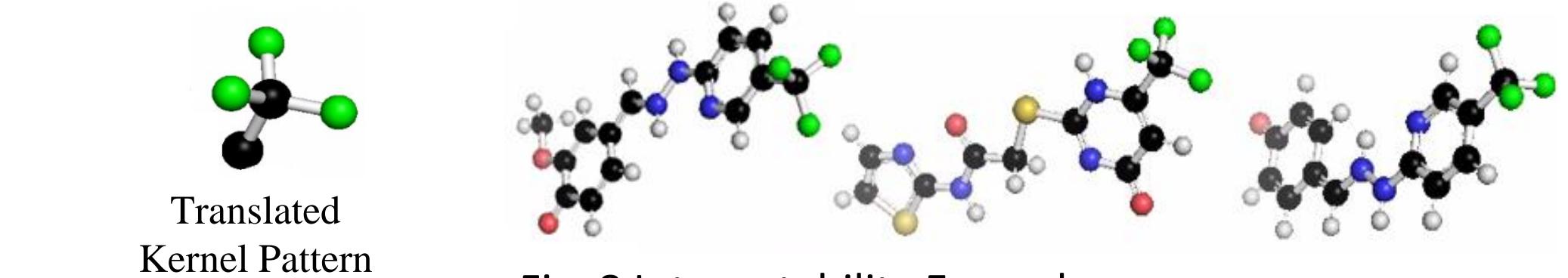


Fig. 8 Interpretability Example

BACKGROUND AND SIGNIFICANCE

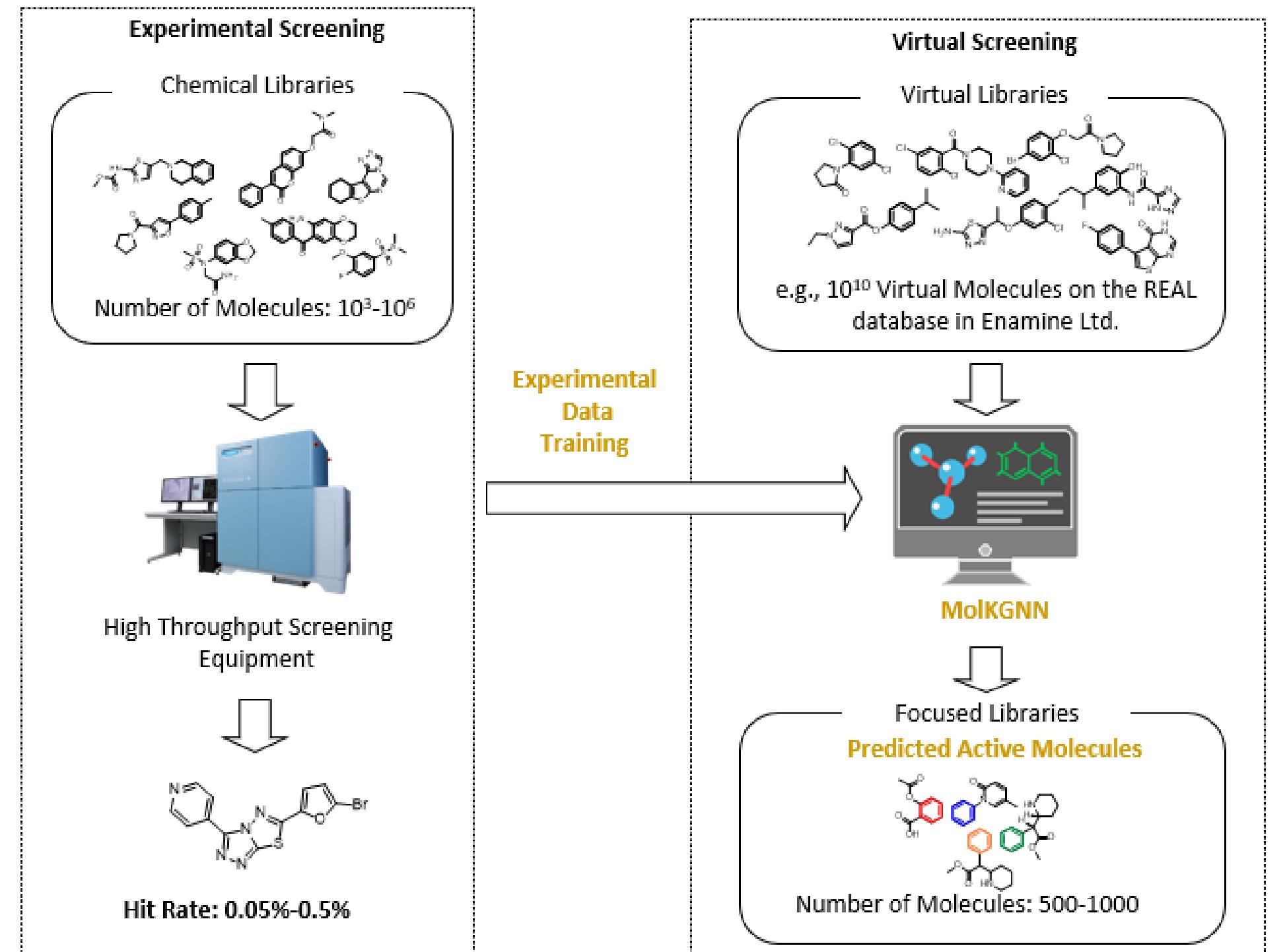


Fig. 2 MolKGNN could enable a larger exploration of the chemical space

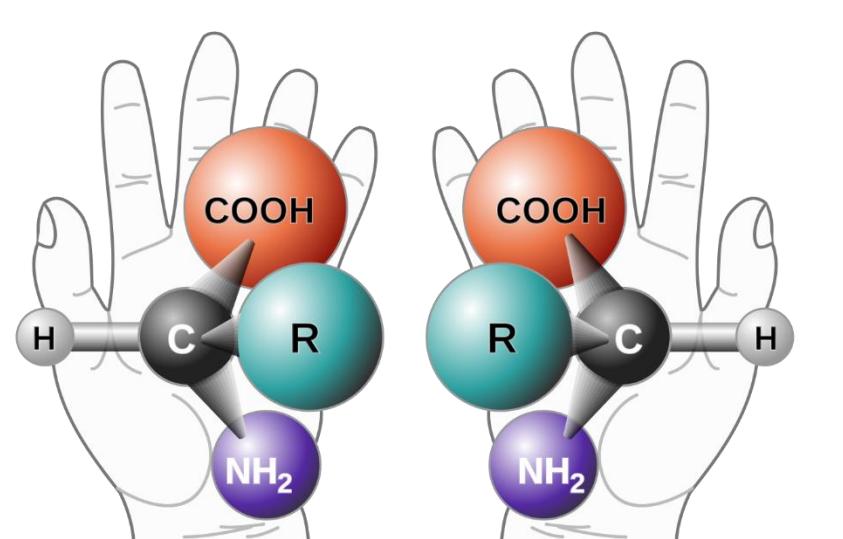


Fig. 3 Challenges: chirality and interpretability

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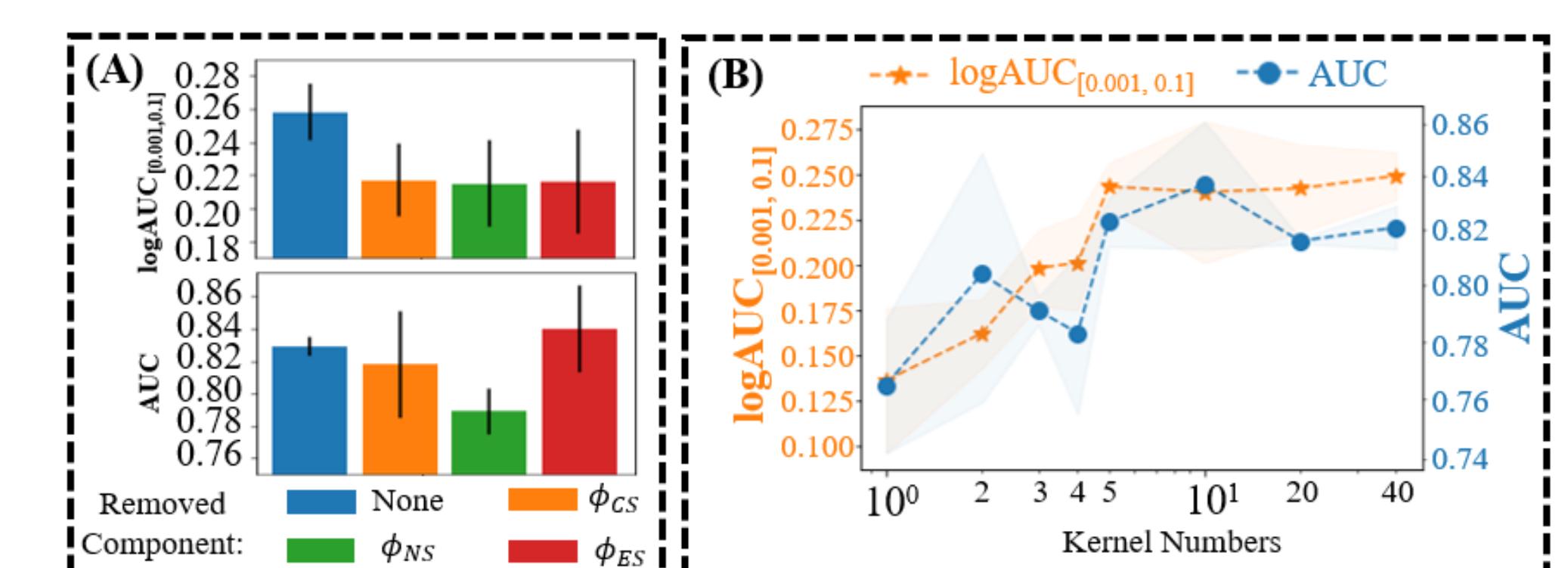


Fig. 9 Ablation Study