

# Quantitative Evaluation of Explainable Graph Neural Networks for Molecular Property Prediction

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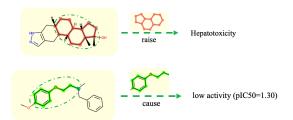
<sup>2</sup>Galixir



# **Background**

#### Motivating question:

- Why do GNN models make the predictions they do?
- Whether the models learn the key substructures relevant for predicting labels correctly?



#### Challenge:

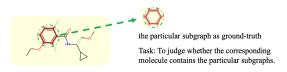
to quantitatively assess the interpretability of GNN

#### Currently Explainable datasets:

Without subgraph ground-truth

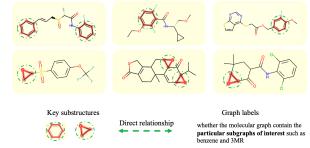


With subgraph ground-truth

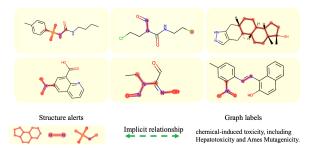


## **Construction of Benchmark**

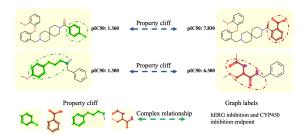
Single rationale: Regular Subgraph



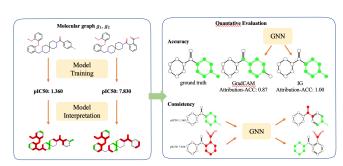
#### Multiple rationales: Diverse subgraph



#### Property cliff: Uncertain subgraph



### A uniform and rigorous Framework



# **Experimental Results**

