Optical Molecular Recognition From Chemical Reaction Mechanism Images



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Background

- Existing OCSR models mainly target properly drawn molecules
- Noises in chemical reaction mechanisms are essential for chemists, but they are structurally similar to the molecule, causing identification errors
- Relying purely on CNN or GNN models for filtering noises is computationally expensive

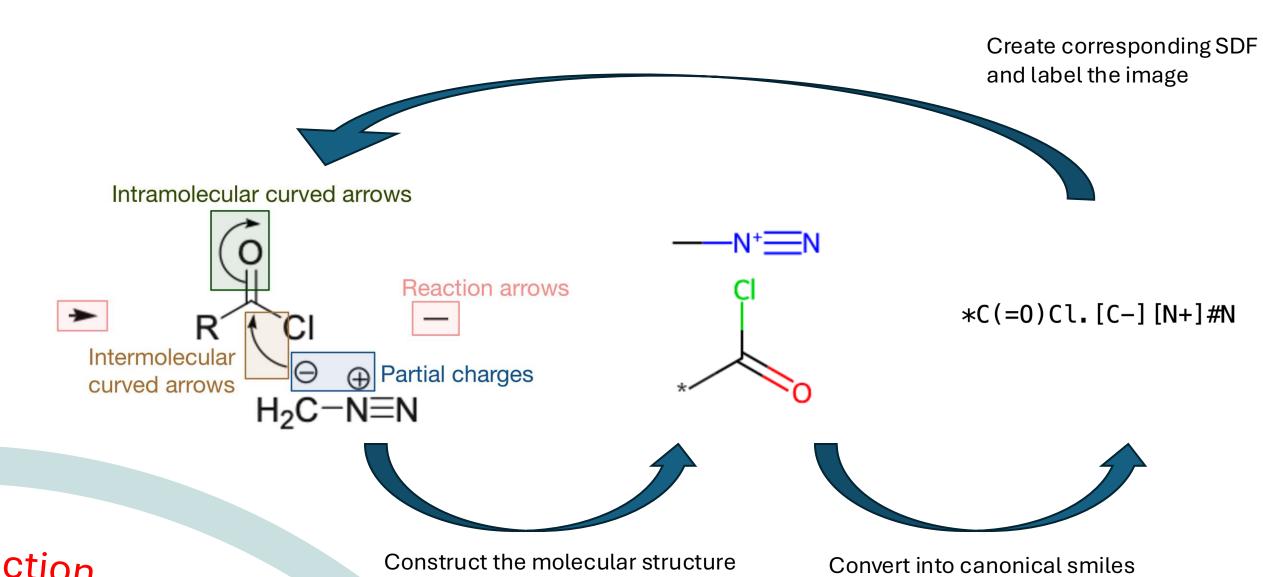
Highlights

- Proposed an image preprocessing technique for optical molecular recognition
- Propose an automated pipeline for processing molecular and reaction information from noisy data
- Created molecular and reaction datasets targeting specifically chemical reaction mechanisms

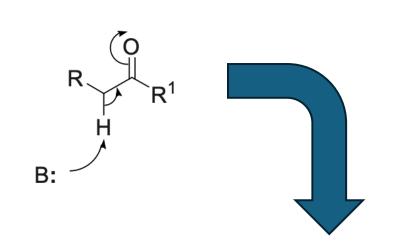
Molecular Dataset

• Created a dataset of 453 molecules from chemical reaction mechanisms [2]

- They include mechanistic features such as curved arrows and partial charges
- Manually annotated with ASKCOS [1] for their SMILES and RDKit for Structural Mol Files



Recognition Pipeline



- Arrow segmentation based on U-net
- Manual annotated 150 images with CVAT for training

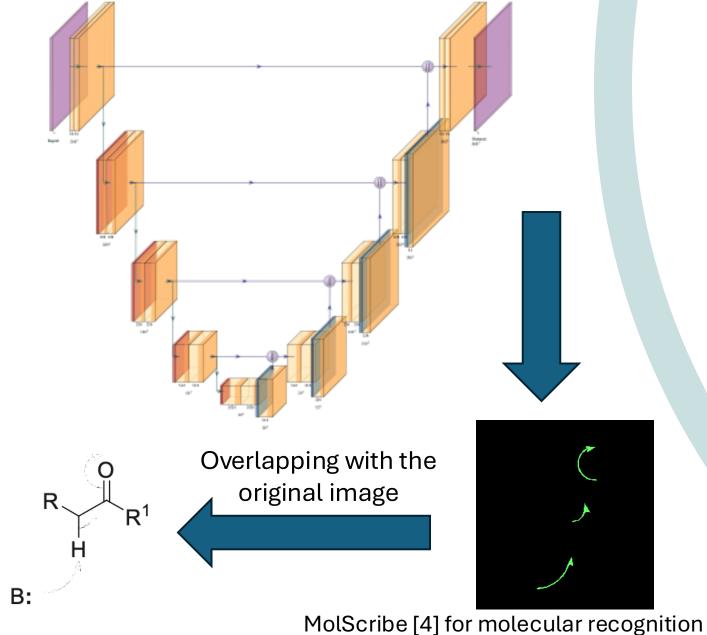


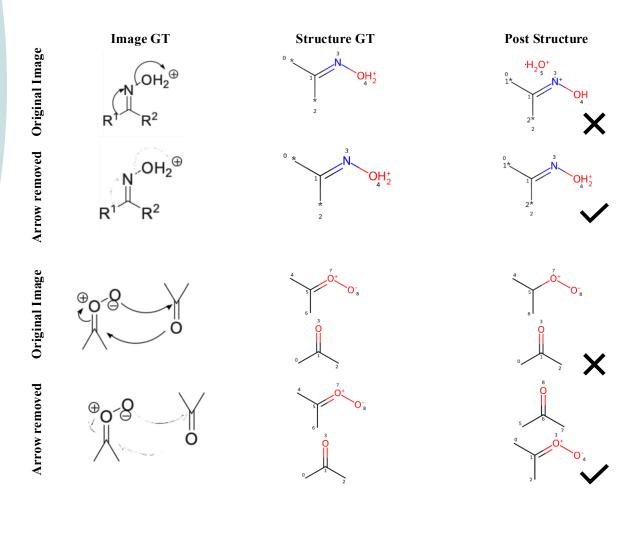
Image Encoder Graph Decoder Transformer Swin Transforme Graph SMILES Data collection

Collect target chemical data from literatures

Optical Molecular Recognition Remove of the contract of the necessary to identify the constitution of the

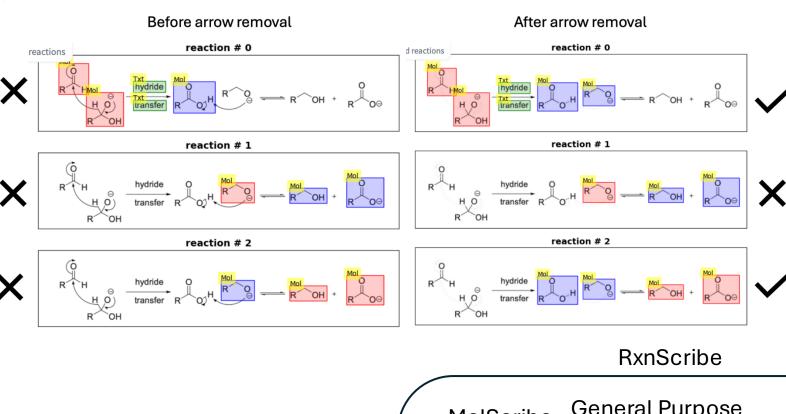
Results

- The presence of curved arrows significantly affects the identification of bonds
- Removal of curved arrows can help retain the essential information of molecule identification while remove noises.



- Shows significant improvement in the used OCSR model evaluation metrics
- Performs as well as their collected dataset
- Perform well in both a sequence-to-sequence model or a CNN encoder scenario

Parsed Reaction Dataset



Further application with increased accuracy on molecular identity and position

[B].*CC(=O)*

RxnScribe [3] demo on chemical reaction mechanism parsing

Prediction

Develop pipeline for a centralized and unified collection of 296 named chemical reaction mechanisms [2]

General Purpose {"Rxn_name.png": | Neural Network Text recognition MolScribe {"reactants": Reaction {"category": "[Mol]", "bbox": [location], Image In "smiles": " "molfile": }], {"category": "[Txt]", "bbox": [location], "text": [] }], "products": [

Conclusion

• Justify the importance of data preprocessing

• Create datasets targeting chemical reaction mechanisms to further benefit both computer scientists and chemists

References

[1] C. Coley et al, "A robotic platform for flow synthesis of organic compounds informed by Al planning," Science, vol. 365, (6453), pp. eaax1566, 2019. DOI: 10.1126/science.aax1566 [2] J. J. Li, Name Reactions A Collection of Detailed Mechanisms and Synthetic Applications. (4th ed.) 2009. DOI: 10.1007/978-3-642-01053-8

[3] Y. Qian et al, "RxnScribe: A Sequence Generation Model for Reaction Diagram Parsing," Journal of Chemical Information and Modeling, vol. 63, (13), pp. 4030-4041, 2023. . DOI: 10.1021/acs.jcim.3c00439 [4] Y. Qian et al, "MolScribe: Robust Molecular Structure Recognition with Image-to-Graph Generation," Journal of Chemical Information and Modeling, vol. 63, (7), pp. 1925–1934, 2023. . DOI:

using transformers," *J Cheminform*, vol. 13, (1), pp. 61, 2021. DOI: 10.1186/s13321-021-00538-8

10.1021/acs.jcim.2c01480 [5] K. Rajan, A. Zielesny and C. Steinbeck, "DECIMER 1.0: deep learning for chemical image recognition

{"category": "[Mol]", "bbox": [location],

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