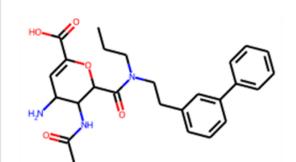


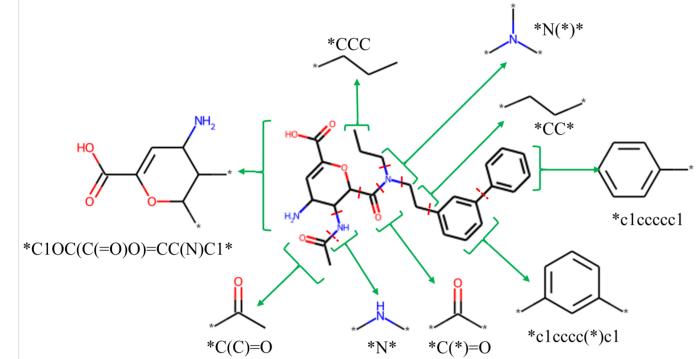
Improving Fragment-Based Deep Molecular Generative Models

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Overview



CCCN(CCc1cccc(-c2ccccc2)c1)C(=O)C1OC(C(=O)O)=CC(N)C1NC(C)=O



- 1. Deep molecular generative models are promising in exploring the molecular space, estimated to be 10^{60} .
- 2. We propose a novel fragmentation and fragment reconstruction algorithm that produces fragments much more efficiently whilst guaranteeing the reconstruction of the original molecule.
- 3. Our model demonstrates superior performance metrics of the generated molecules by simply applying a factor to the sampling variance rather than using low-frequency masking [1].



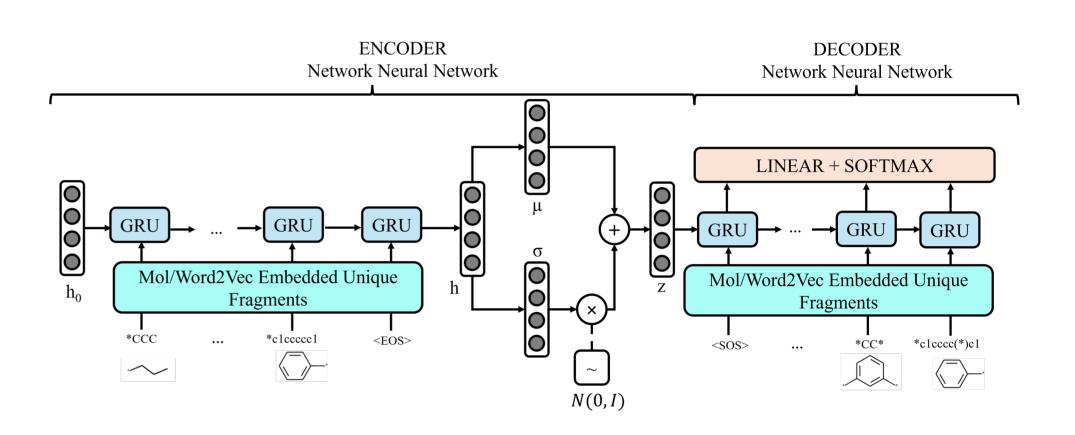
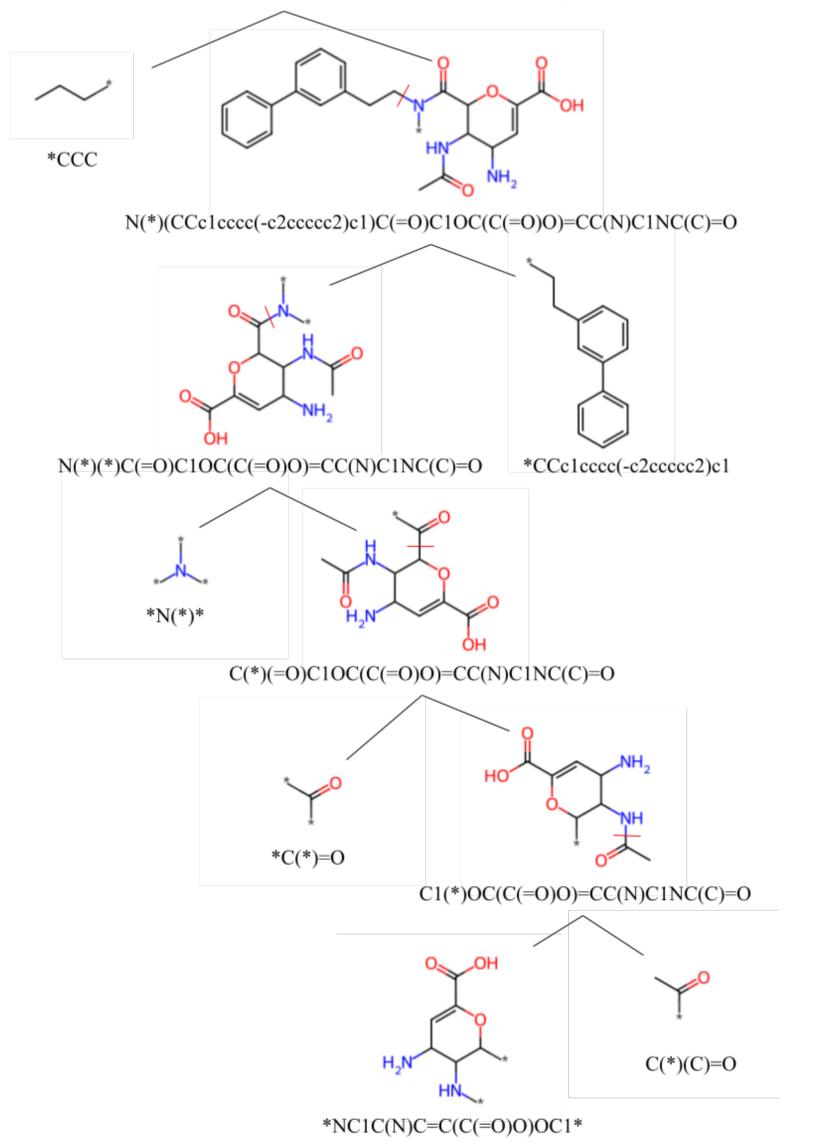


Figure 1: Example of fragments derived from a molecule using the BRICS algorithm. The asterisks represent dummy atoms, and the red dotted lines denote the 'cuts' in the molecule.

CCCN(CCc1cccc(-c2ccccc2)c1)C(=O)C1OC(C(=O)O)=CC(N)C1NC(C)=O



Results

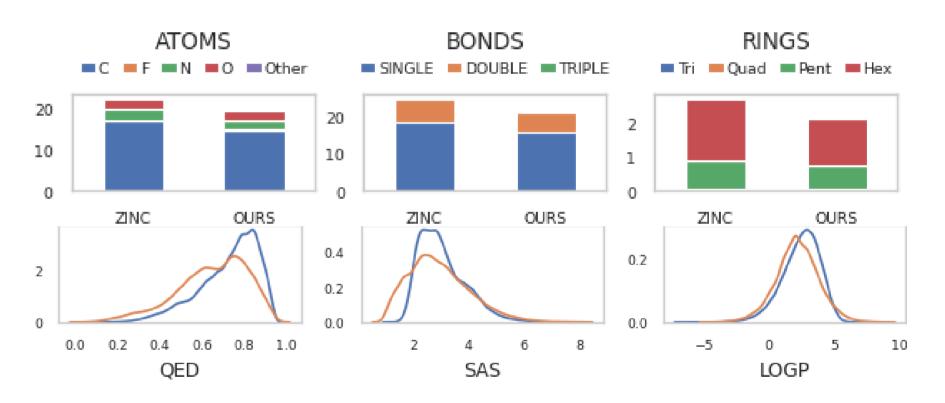


Figure 3: Top: Statistics of sampled molecular characteristics. Bottom: Density of sampled molecular properties where the blue line corresponds to ZINC. Our model: c = 25, Mol2Vec Embedding.

Model	Valid	Novel	Unique Sample Rate (ms/mol)			
Word2Vec Embedding						
Podda	1	0.992	0.460			
Mol2Vec Embedding						

Ours (c = 20)	1	0.997	0.661	123			
Ours (c = 25)	1	0.997	0.793	211			
Low Frequency Mask Sampling							

Podda (LFM) 0.995 0.998

Table 1: Results of 20,000 molecules sampled

References

Figure 2: Example of fragments derived from a molecule using Breadth-First Fragmentation algorithm. The asterisks represent dummy atoms, and the red dotted lines denote the 'cuts' in each recursion.

[1] Marco Podda, Davide Bacciu, and Alessio Micheli. A Deep Generative Model for Fragment-Based Molecule Generation. In Proceedings of the Twenty Third International Conference on Artificial Intelligence and Statistics, pages 2240–2250. PMLR, June 2020. ISSN: 2640-3498.