This project involved the development of a computer protocol using the open-source chemical library RDKit and Python as the programming language. From the structure of



a chemical backbone (Pyridinophane macrocycles), the protocol generates different derivative 2D structures by varying the attaching groups and their positions on the backbone structure, up to 18,000 final structures. The protocol then screens these structures for many desired chemical and pharmacological properties in order to narrow down the best few candidates for synthesis. This protocol can save time and resources for synthesis labs while providing a way to optimize chemical backbones in drug design.

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### INTRODUCTION

Oxidative stress refers to the imbalance between free radical activities and antioxidant activities in the body, and is known to be involved in the development of various neurodegenerative diseases (Alzheimer's and Parkinson's). To help the body target and rebalance this process, the Green group at TCU has designed pyridinophane macrocycle frameworks (PyN<sub>3</sub>,  $Py_2N_2$ ) for the development of a small multimodal molecule with direct targeting of oxidative stress. To enhance antioxidant activity, the group proposed a library of ligands as modifications to the pyridinophane frameworks.

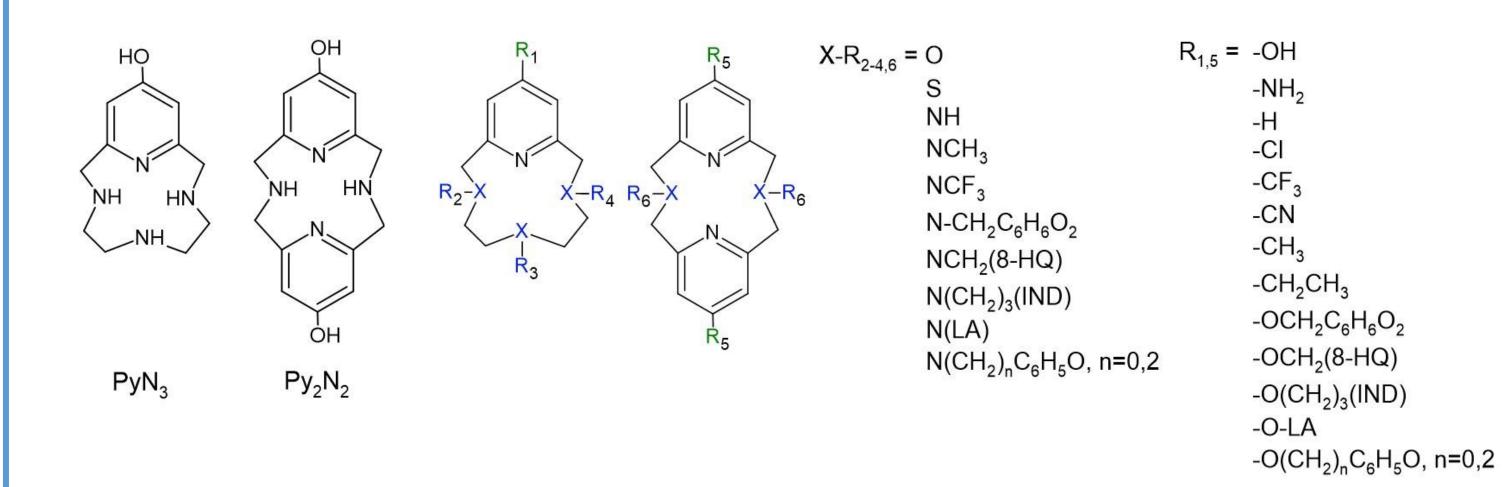


Figure 1. TCU Green group's successfully synthesized pyridinophane macrocycle frameworks (PyN<sub>3</sub> and Py<sub>2</sub>N<sub>2</sub>) and their proposed library of ligands as modifications to the backbone structures.

Due to the resulting large number of possible structures, computational pre-screening is essential to select the most promising candidates for synthesis as well as for biological and chemical testing. The ligand screening utilized the RDKit open-source cheminformatics toolkit,<sup>1</sup> with Python3 interface. The developed protocol is capable of generating the **2D** structures of all the possible modified structures of the pyridinophane frameworks and screen them for "drug-likeness" based on Lipinski's "Rule of Five." The properties computed and used as screening conditions in this project were molecular weight (MW), ring count, octanol-water partition coefficient (cLogP), number of hydrogen bond donors (HBD) and acceptors (HBA), and polar surface area (PSA).

Number of possible modified structures:

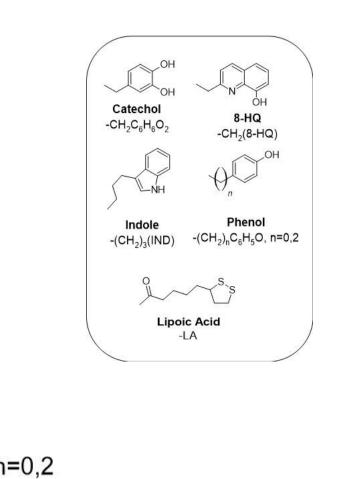
 $PyN_3$ : 11<sup>3</sup> x 14 = **18,634**  $Py_2N_2$ : 11 x 14 = **154** 

Screening Protocol:  $MW \le 500$  $cLogP \le 5.0$  $\text{HBA} \le 10$  $HBD \leq 5$  $PSA \le 90$ 

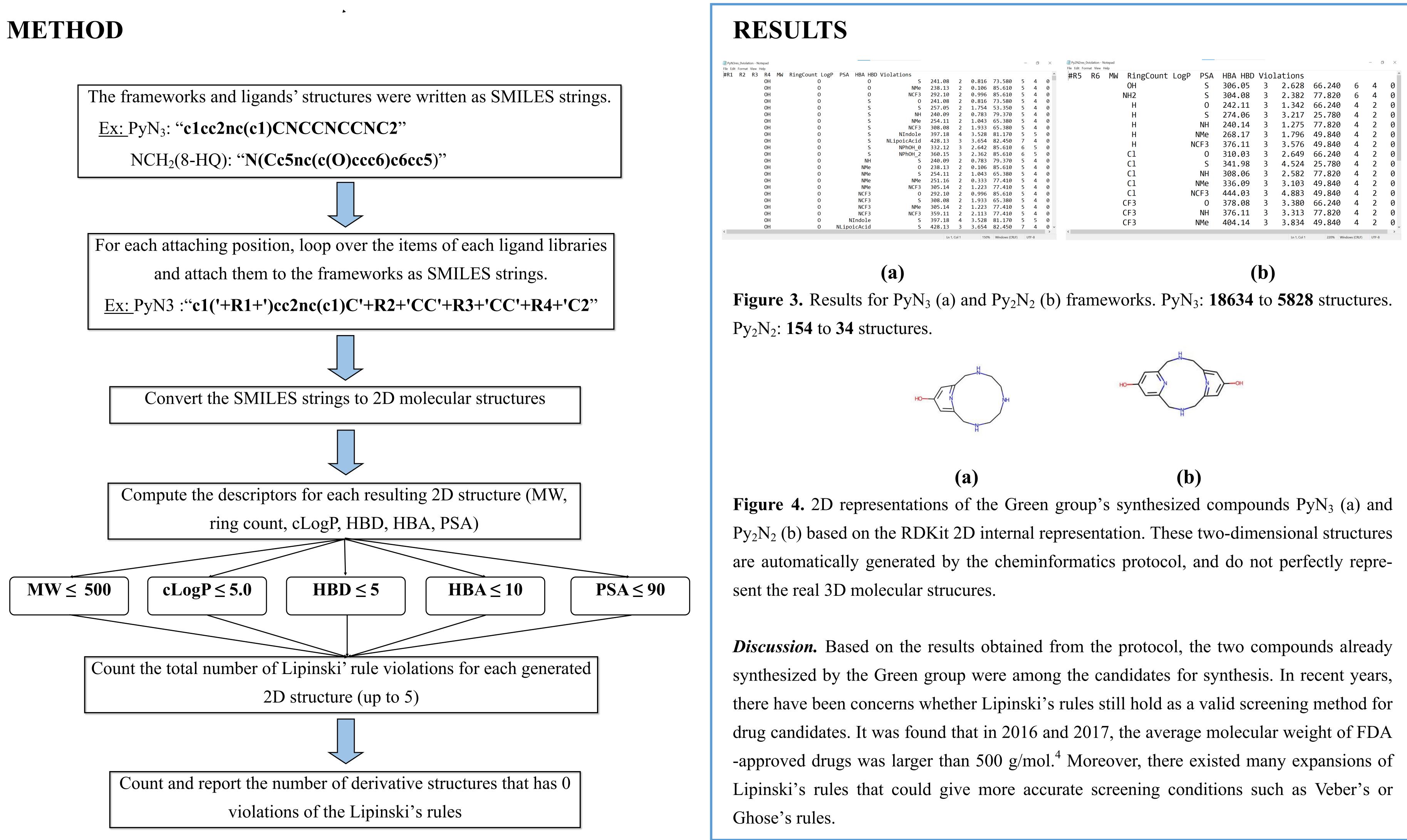
**Figure 2.** A general scheme of the screening protocol.

# **Development of a Virtual Screening Protocol for Pyridinophane Macrocycle Derivatives as Therapies for Oxidative Stress**

### Minh Nhat Pham, Benjamin G. Janesko\*



PyN<sub>3</sub>: **5828** Py<sub>2</sub>N<sub>2</sub>: **34** 



#### **Lipinski's Rule of Five:**

In 1977, after assessing various approved drugs and clinical candidates at that time, Lipinski and co-workers proposed the "Rule of Five" for orally active compounds. For a drug molecule to be orally bioavailable, the molecule's lipophilicity and solubility are important factors in intestinal absorption. Molecules that are small (MW  $\leq$ 500), less soluble in water (cLogP  $\leq$  5.0), has small total polar surface area (PSA  $\leq$ 90) and hydrogen bond counts (HBD  $\leq$  5 and HBA  $\leq$  10) are more likely to cross the lipid bilayers of the cell membrane.<sup>2</sup> Molecules failing to comply these rules may show poor absorption, faster rate of metabolism and excretion, unfavorable distribution, and toxicity.<sup>3</sup>

### CONCLUSION

From the starting 18,634 possible structures for the derivative of the pyridinophane framework PyN<sub>3</sub>, the developed protocol were able to narrow the list of candidates to 5828 structures with 0 violations of the Lipinski's rules. Similarly, the Py<sub>2</sub>N<sub>2</sub> framework also yielded 34 best candidates from the original 154 options. In the next stage, 3D structures of these compounds can be generated and subjected to DFT (density functional theory) calculations for more advanced properties such as metal binding ability, pKa and pI, HAT, and redox potentials to further reduce the list to 20 - 30 compounds.

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