## Improved Generation of Symmetric Cyclic Peptide Complexes with RFDiffusion

**Rationale:** Cyclic peptides (*Figure 1*) have demonstrated promising potential for their structural stability, membrane and cell permeability, and high binding affinity in the field of pharmaceuticals and drug design. However, the *de novo* design of such molecules has been (A) mostly limited to classical, energy-based design techniques and (B) focused on the design of cyclic peptides as monomers (that then associate with targets of interest), rather than as homo-oligomeric species.

Symmetric structures in cyclic peptides present an exciting area of study for several reasons. First, they allow for the robust generation of new particles, expanding the possibilities for molecular design. Second, they can access a broader range of length scales, providing more flexibility in designing molecules that fit specific binding targets. Lastly, studying symmetric structures serves as a stepping stone toward understanding multi-component oligomers.

Recently, a generative neural network, RFDiffusion, has demonstrated the successful generation of

protein oligomers with high experimental success rates.<sup>3</sup> Furthermore, recent findings have shown that manipulating the relative positional encoding of amino acids can generate cyclized peptide predictions using RFDiffusion.<sup>4</sup> Leveraging these findings, we aim to combine the symmetric oligomer generation capability of the RFDiffusion model with that of producing cyclic peptides and explore their experimental feasibility.

Then, we will validate the robustness of those generated symmetric oligomers through comparative analysis with experimentally determined structures. This will be a cornerstone for the efficient generation of symmetric conformation of macrocyclic polypeptides.<sup>4</sup>

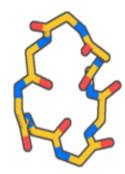


Figure 1. PyMOL Implementation of a cyclic peptide.

## **Hypotheses:**

- We can make symmetric oligomers of cyclic peptides, and they maintain cyclic peptide properties such as cell permeability, increased binding affinity, and structural stability.
- We can gain more insights and control over the designing process for *de novo* protein binders and other targets.

**Acknowledgment:** First and foremost, I would express my deepest gratitude to **David C. Juergens** for providing insightful guidance and professional mentorship to proceed with this research. In addition to his immense support, I appreciate **Dr. Gaurav** for initiating cyclic peptide prediction research which is a main source of initiating this project.

**Objective:** Robust generation process of symmetric oligomers made of cyclic peptides.

**General Approach:** Our main approach for the robust generation and testing of symmetric oligomers of cyclic peptides will include the following key steps.

- We will integrate the protocols in RFDiffusion for generating symmetric oligomers with its capacity to produce cyclized peptides (*Figure 2*). This combination aims to harness the structural stability and high binding affinity of cyclic peptides.
- We can use cyclized peptides as an asymmetric unit. Then, we can generate complexes with diverse point group symmetries.
- To design the sequences for these RFDiffusion backbone structures, we will use ProteinMPNN.<sup>5</sup> This model will ensure that our generated structures have sequences with high performance and structural compatibility.
- We will use cyclic AlphaFold2 (AF2) structure prediction protocol to filter designs in silico and select designs for experimental validations.<sup>1</sup>
- We will perform a comparative analysis using *r.m.s.d* and confidence metrics for validating the robustness of generated structures.

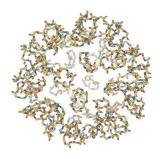


Figure 2. PyMOL Implementation of Symmetric oligomers designed with cyclized peptides.

## References

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