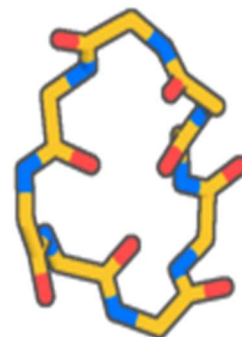


# Improved Generation of Symmetric Cyclic Peptide Complexes with RFDiffusion

**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.<sup>1</sup>

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. Recently, a generative neural network, RFDiffusion, has demonstrated the successful generation of protein oligomers with high experimental success rates.<sup>2</sup>

Furthermore, recent findings have shown that manipulating the relative positional encoding of amino acids can generate cyclized peptide predictions using RFDiffusion.<sup>3</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.



*Figure 1. PyMOL Implementation of a cyclic peptide.*

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>3</sup>

## Hypotheses:

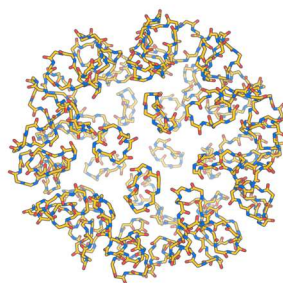
- We can make symmetric oligomers of cyclic peptides, and they maintain cyclic peptide properties such as cell permeability, increased binding affinity, and structural stability.
- We can gain more insights and control over the designing process for *de novo* protein binders and other targets.

**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>4</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.*

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