

# Predicting Peptide Structural Feasibility with CNNs

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## Project Goal

Predict the structural feasibility of peptides utilizing CNNs

## Task 1: Training Set Creation

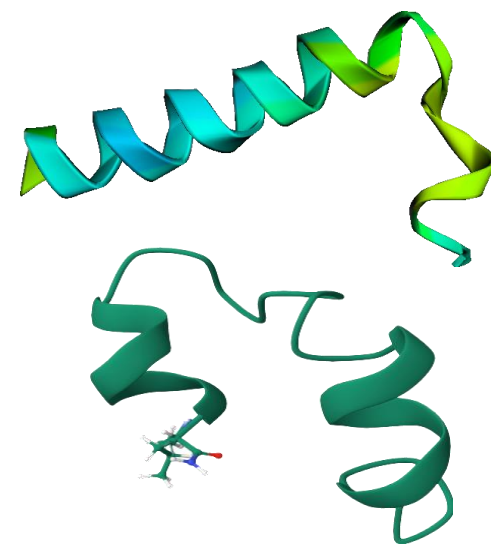
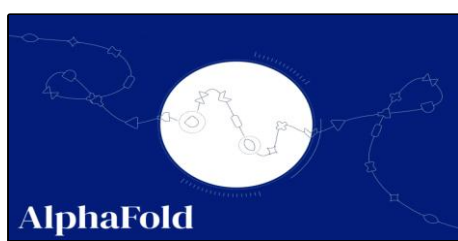
Create a labeled training set by comparing the structural similarity of generated peptides to their real counterparts

## Motivation

Provide researchers with a predictive tool to accelerate the analysis of peptide model stability.

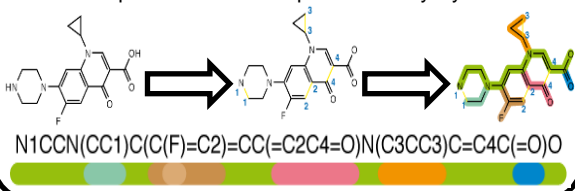
Sequence:

SPLGEEMRDRARAHVDALRTHLAPYS  
DELRQRLAARLEALKENGGA



## S.M.I.L.E.S

Simplified Molecular-Input Line-Entry System



Adapt Data

Focus:

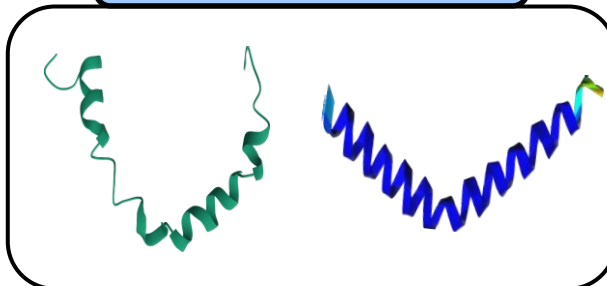
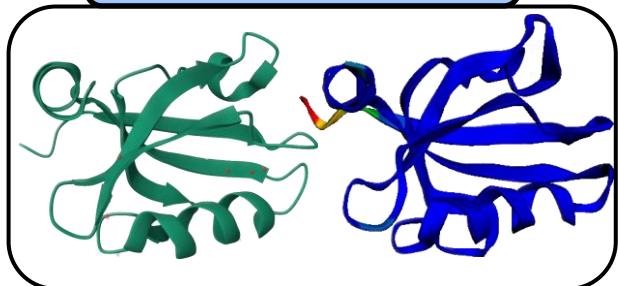
- Reduce Dimensionality
- Nonadjacent bonding
- Structural components

.PDB Format

ATOM	1	N	HIS	B	-1	75.822	-55.481	-8.354	1.00	47.79
ATOM	2	CA	HIS	B	-1	76.394	-55.280	-7.029	1.00	57.91
ATOM	3	C	HIS	B	-1	76.626	-53.799	-6.752	1.00	52.10
ATOM	4	O	HIS	B	-1	77.492	-53.435	-5.956	1.00	47.84
ATOM	5	CB	HIS	B	-1	77.707	-56.053	-6.892	1.00	54.37

Positive Label: Similar Structure

Negative Label: Dissimilar Structure



Compare SMILES for Similarity

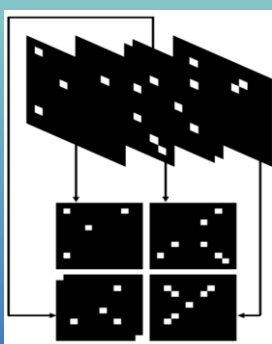
## Preliminary Results

Peptide ID:	Cosine Similarity:	Label
Shu3	0.9974	Positive
N7I	0.9913	Positive
e0a	0.2074	Negative
M8I	0.2635	Negative
1CEU	0.3680	Negative
1Hzn	0.4200	Negative
Gw4	0.1675	Negative
1Loi	0.1848	Negative

## Future Key Tasks

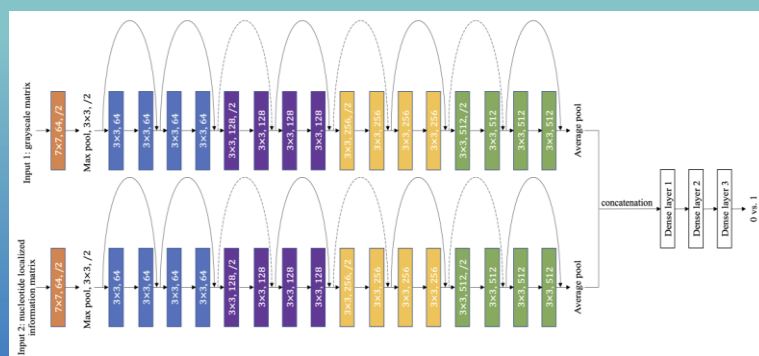
### Task 2: Adapt NU-Resnet for Peptide Sequences

Adapt current RNA based CNN to take in and predict peptide sequences



### Task 3: Adapt NUMO-Resnet to utilize Peptide Motifs

Implement motif input data for NUMO-Resnet CNN to utilize



## Sources

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- SMILES, a chemical language and information system. 1. Introduction to methodology and encoding rules David Weininger *Journal of Chemical Information and Computer Sciences* 1988 28 (1), 31-36 DOI: 10.1021/ci00057a005

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