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# **Protein Loop Modeling Using Deep Generative Adversarial Network (GAN)**

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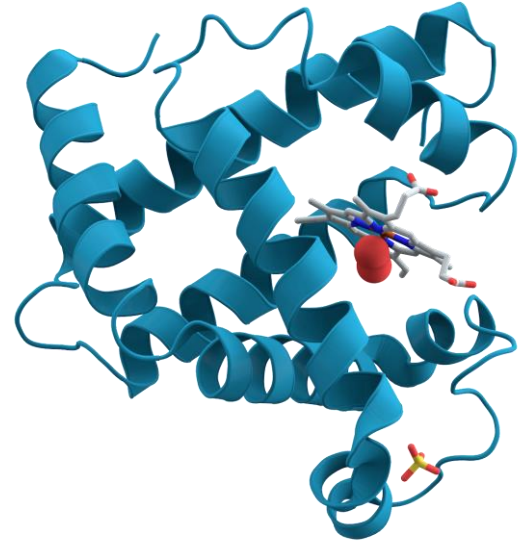
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  - First successful application of GAN in Bioinformatics
- Background and Motivation
- Methods
  - A New GAN Deep Neural Network for Loop Modeling
- Experimental Results
- Conclusion



# Introduction

Invitation to the world of Bioinformatics

- Proteins are most common molecules in cells
- Important role in medicine and life science
- 20 amino acid types
- Amino acid sequence determines the 3D structure
- During this talk, your body will produce 10,000,000,000,000,000,000 protein molecules

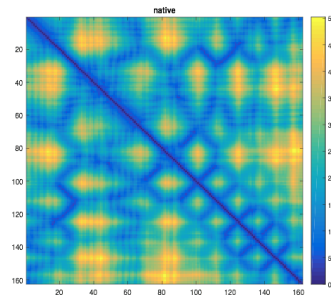


A visualization of the 3D structure of a protein

# Background

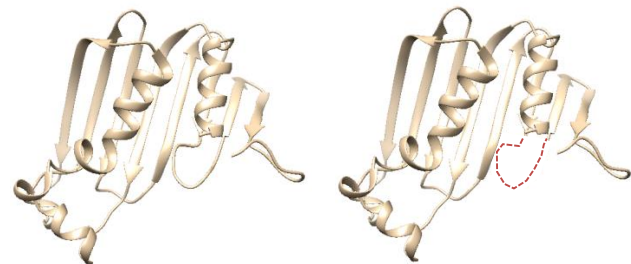
## Distance Map and Multidimensional Scaling (MDS)

- Each amino acid has a coordinate
- Distance map: Euclidean distance of  $C\alpha$  atoms of all amino acid pairs
  - From 3D to 2D
  - Can be treated as an image
  - Orientation independent
- MDS can restore 3D space while preserving the distance constraints in the 2D space



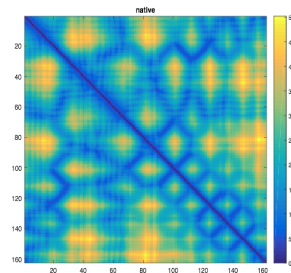
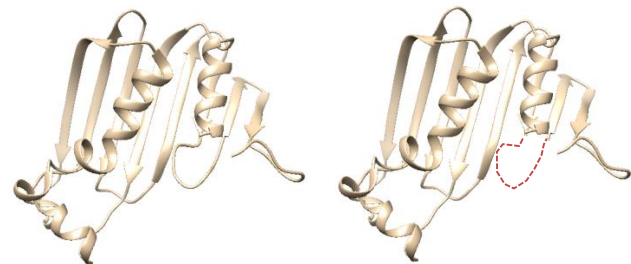
# Protein Loop Modeling

- Small parts in a protein may be missing
- Loop modeling is to predict those missing regions

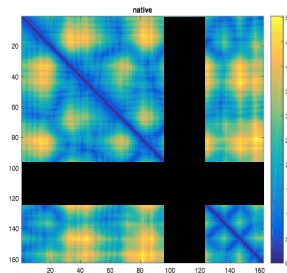


# Protein Loop Modeling

- Small parts in a protein may be missing
- Loop modeling is to predict those missing regions
- We solve this problem in 2D space using distance map



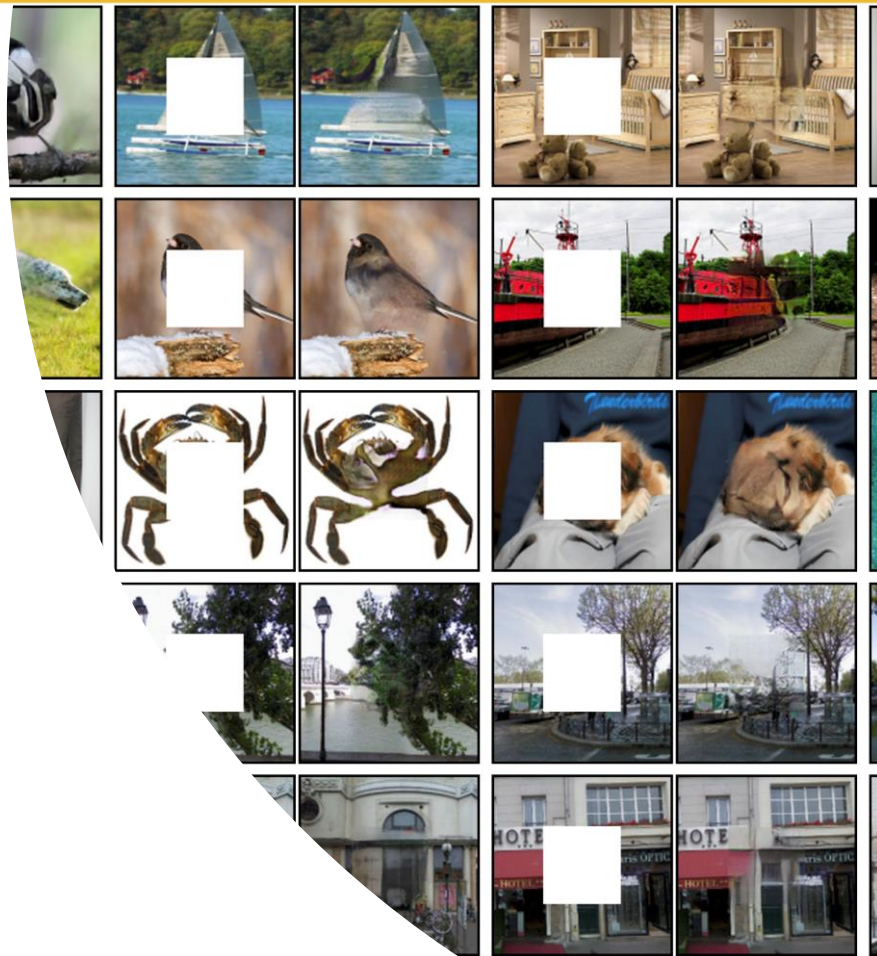
Complete protein



A missing region in  
the protein

# Motivation

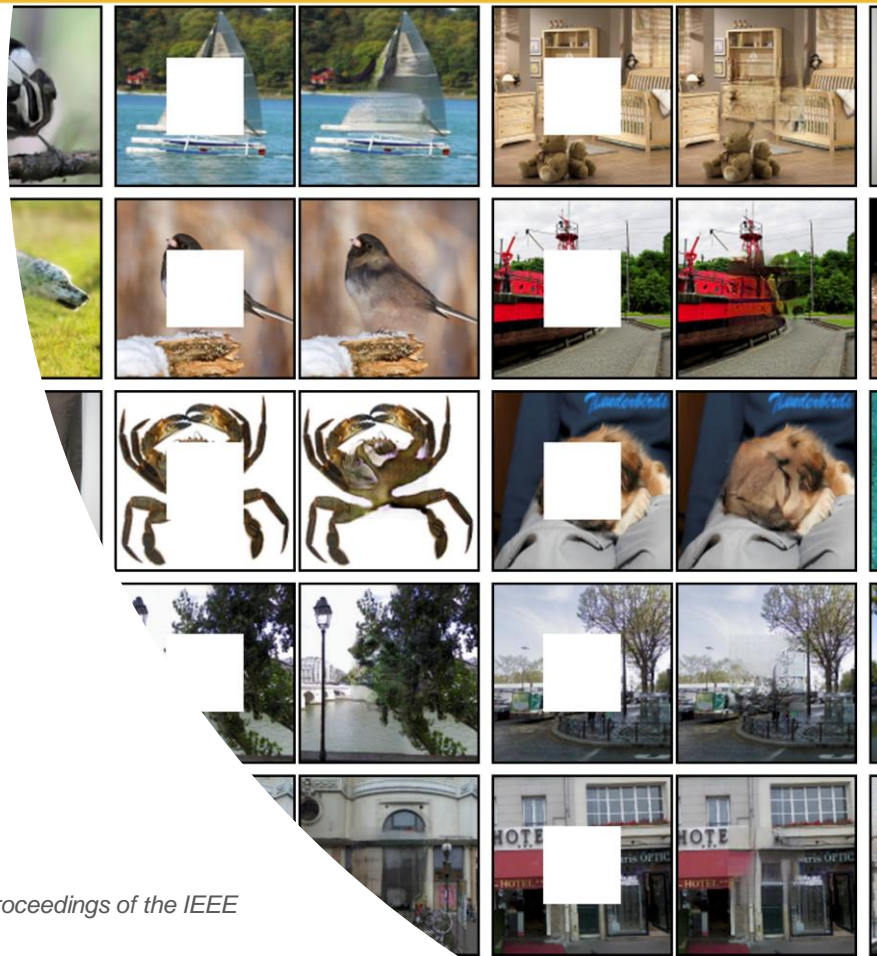
- Image completion (inpainting):  
complete the missing region  
based on the image context





# Motivation

- Image completion (inpainting): complete the missing region based on the image context
- Generative Adversarial Network (GAN) outperformed previous methods [1]

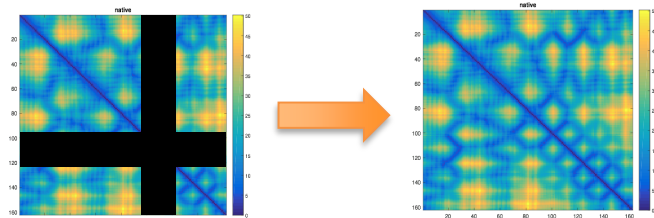


[1] Pathak, Deepak, et al. "Context encoders: Feature learning by inpainting." *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*. 2016.



# Motivation

- Image completion (inpainting): complete the missing region based on the image context
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# Prediction Flow of Our New Method

**Target:**

Length 50 with missing  
region in the middle

CDWEEISVKGPNGESSVIHDR**KSGKKFSI**EEALQSGRLTPAHYDRYVNKD



# Prediction Flow of Our New Method

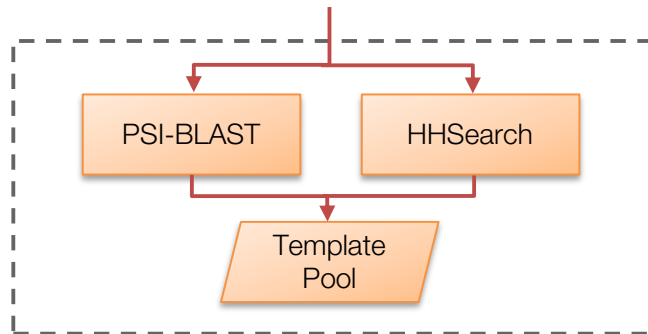
## Target:

Length 50 with missing  
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CDWEEISVKGPNGESSVIHDR**KSGKKFSI**EEALQSGRLTPAHYDRYVNKD

## Alignment Searching:

Templates that can cover the  
whole query target



# Prediction Flow of Our New Method

## Target:

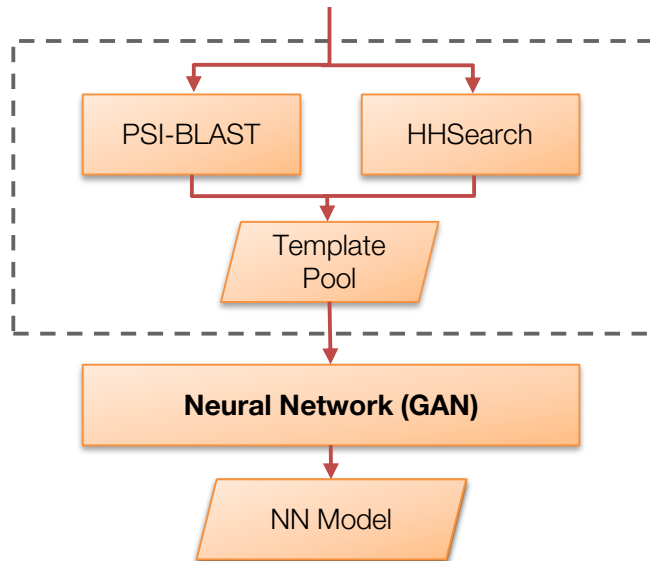
Length 50 with missing  
region in the middle

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## Neural Network Training:



# Prediction Flow of Our New Method

## Target:

Length 50 with missing region in the middle

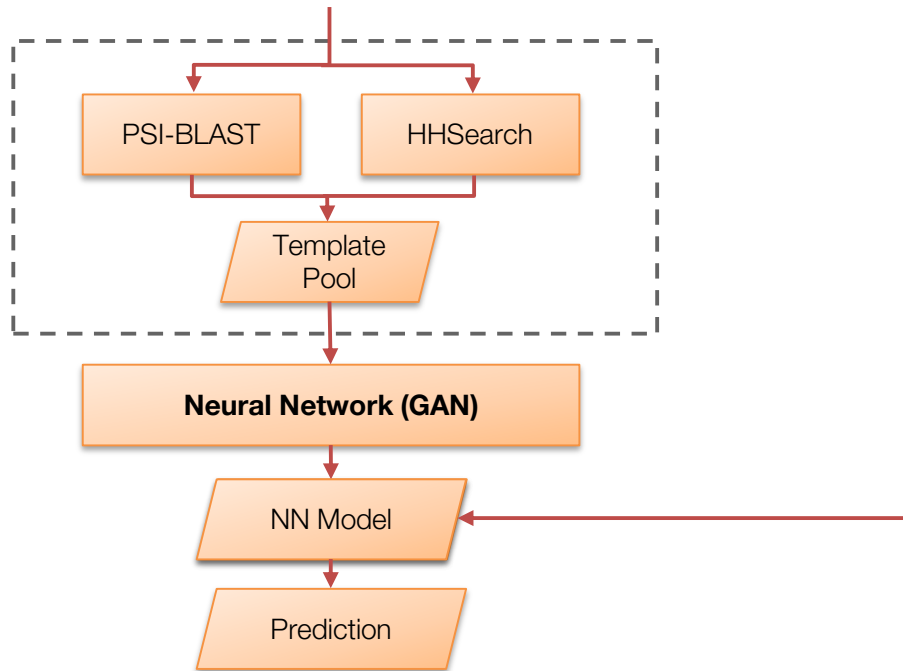
## Alignment Searching:

Templates that can cover the whole query target

## Neural Network Training:

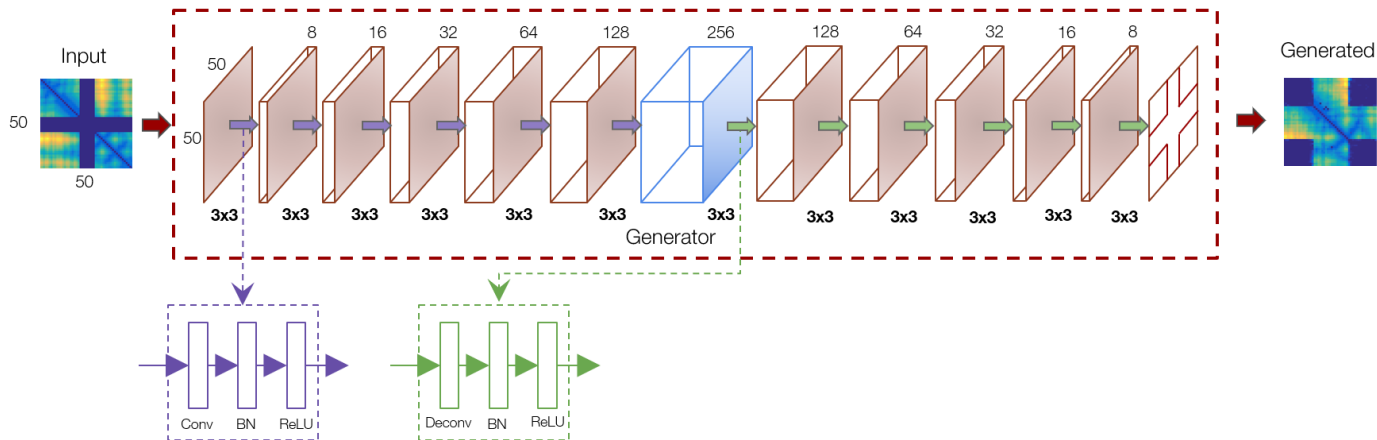
## Prediction:

CDWEEISVKGPNGESSVIHDR**KSGKKFSI**EEALQSGRLTPAHYDRYVNKD



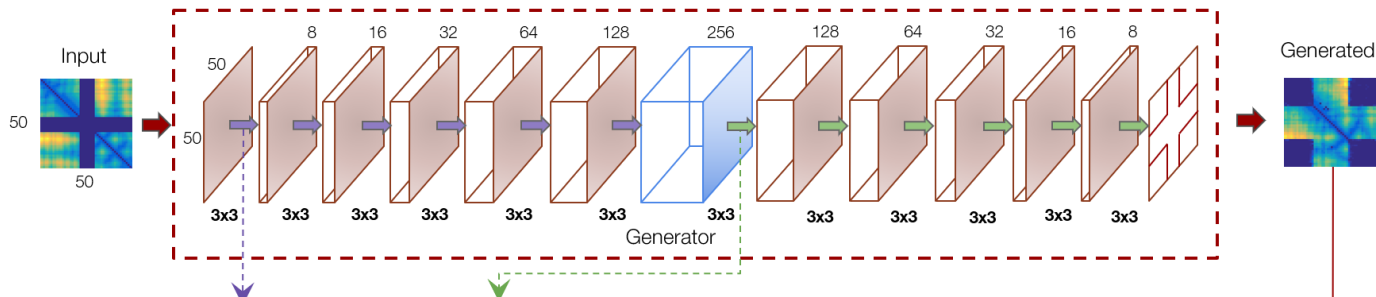
# Our GAN Network Structure

Generator

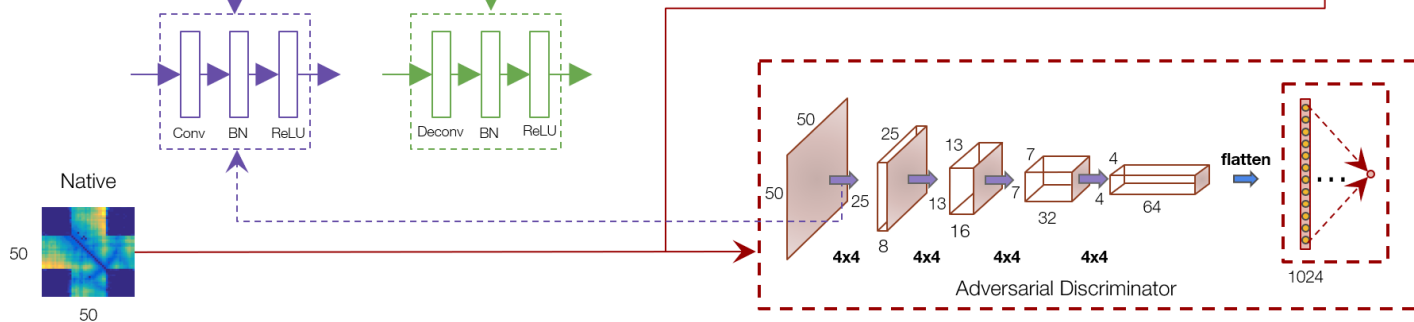


# Our GAN Network Structure

Generator

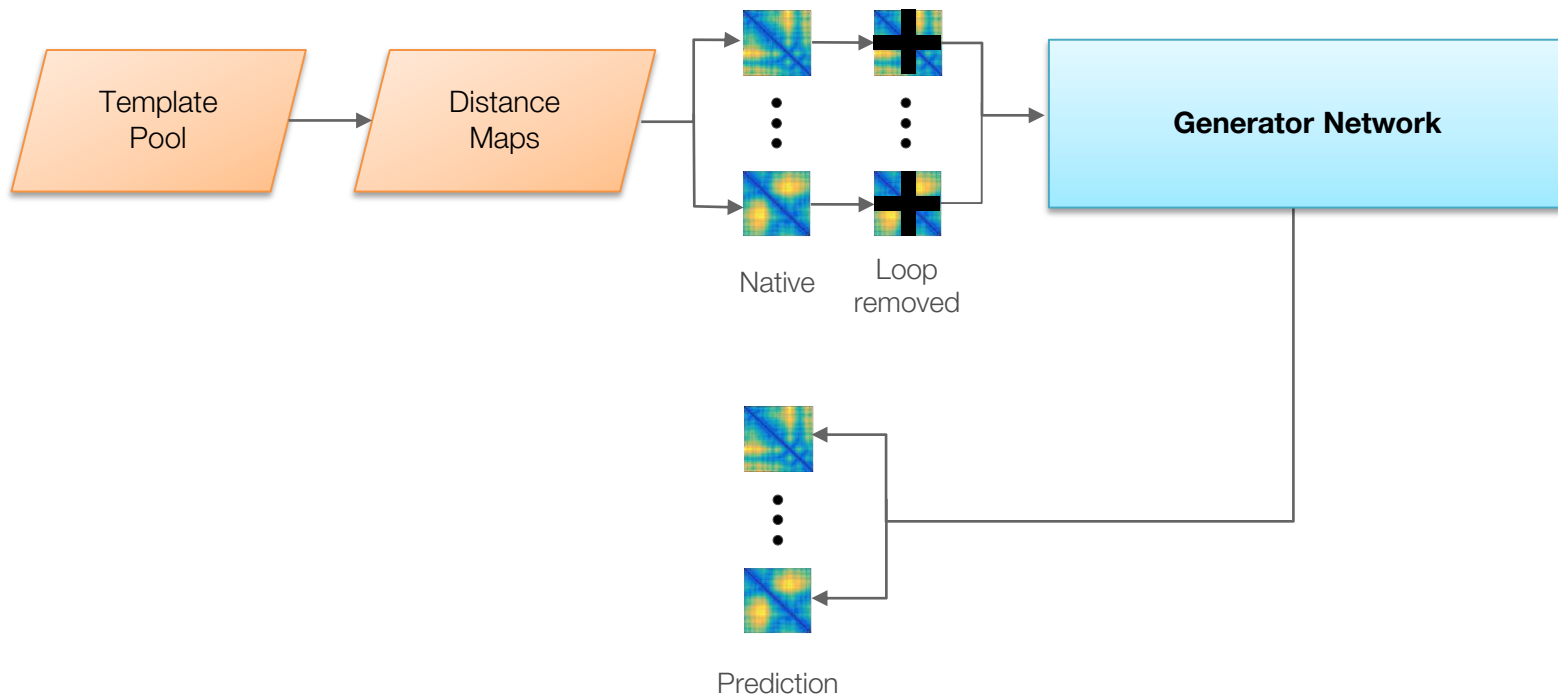


Discriminator

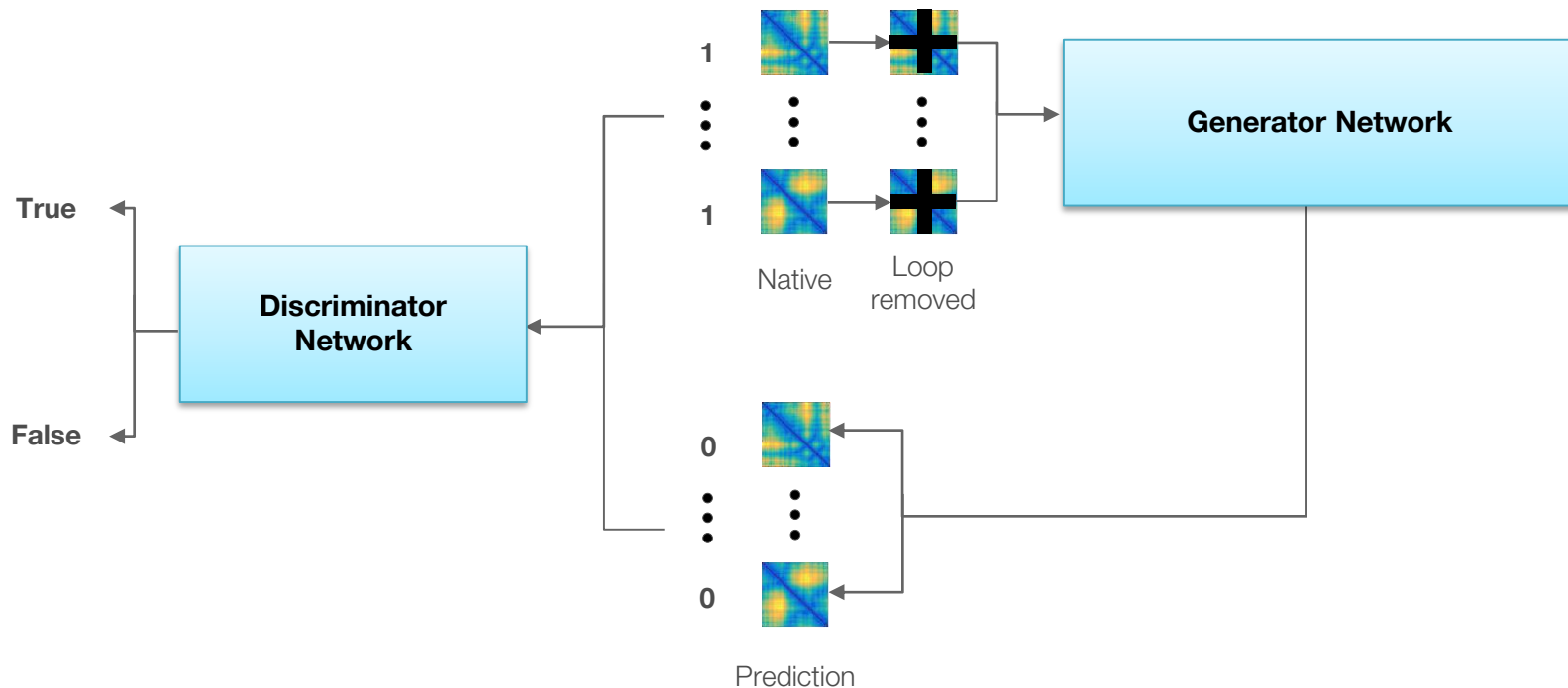




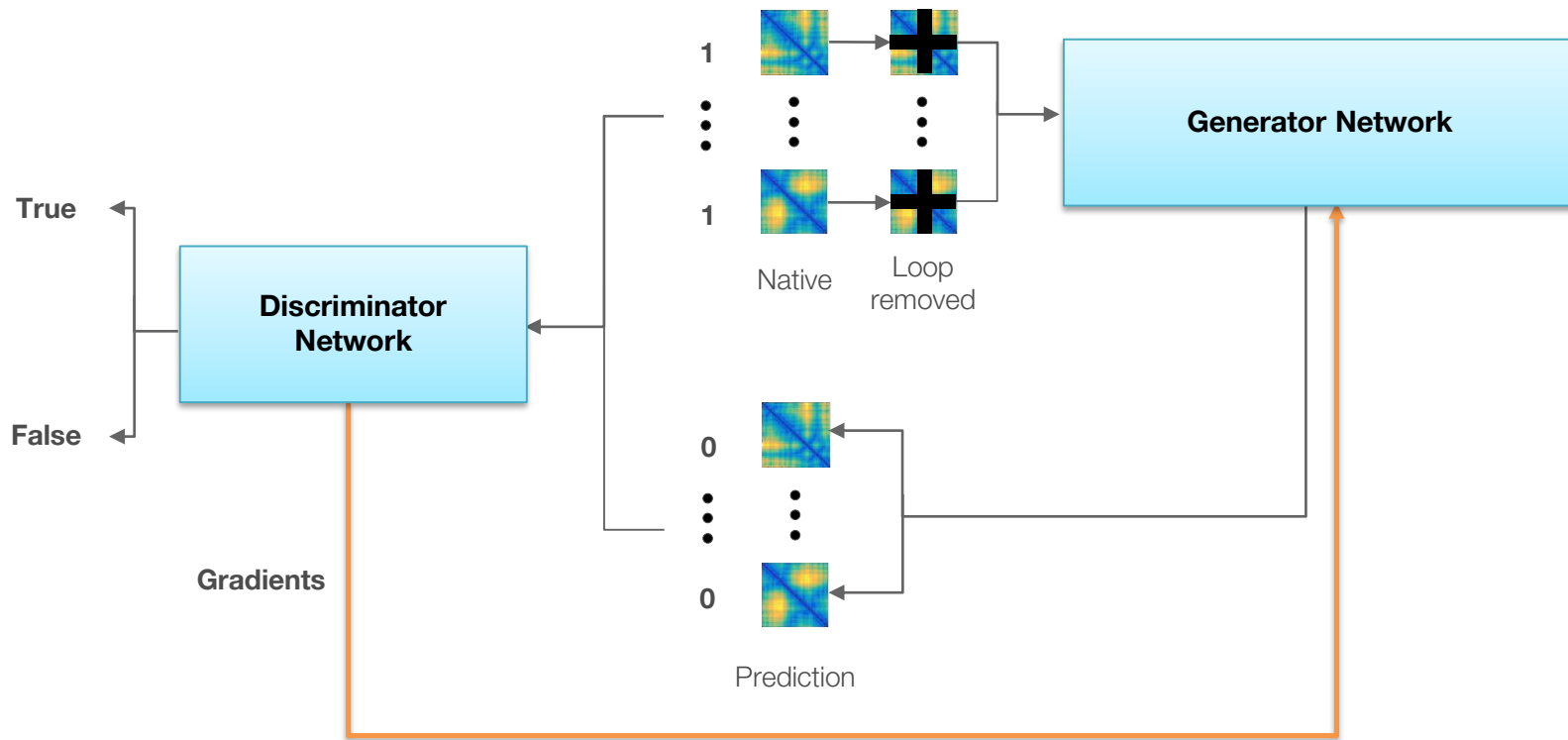
# Training of the Network



# Training of the Network



# Training of the Network



# Implementation Details

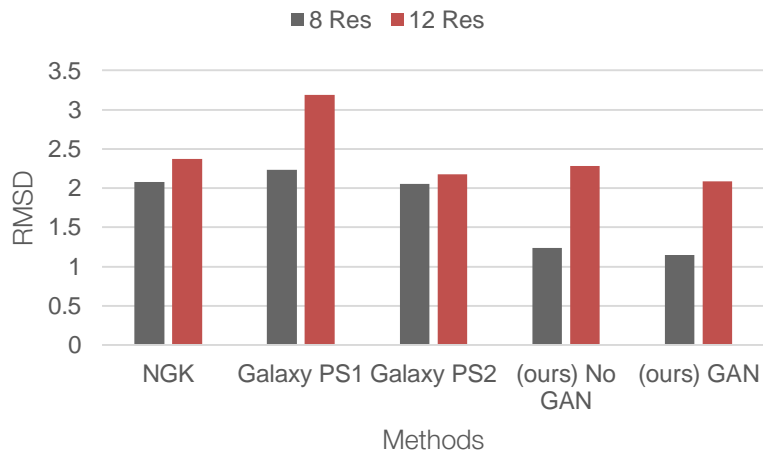
- Platform
  - TensorFlow 1.0
- Training
  - Generator and Discriminator are trained alternatively
  - Train the Generator every step and train Discriminator every 10 steps
- Configuration
  - Adam optimizer with the learning rate 0.0001
  - Early stopping



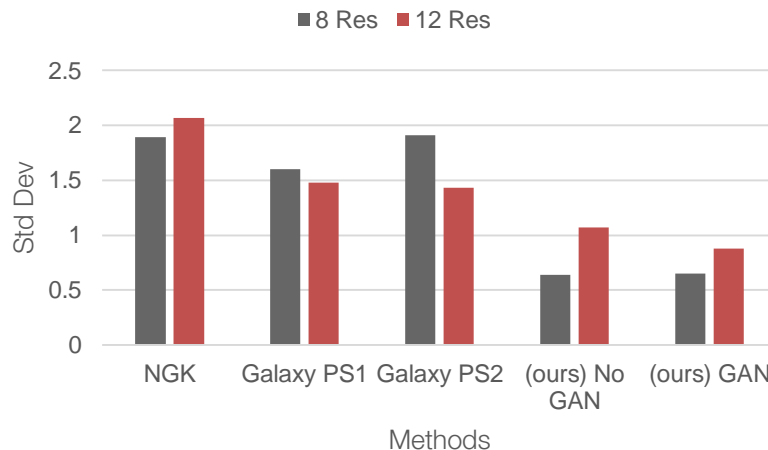
# Experimental Results

- *Without GAN (Generator)* and *with GAN (Generator + Discriminator)*
- Benchmark datasets from Park et al. including 20 targets with 8-length loop (8 Res), 20 with 12-length loop (12 Res)

Average RMSD (smaller is better)

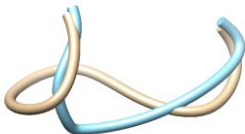

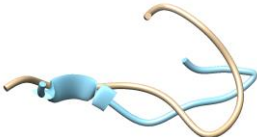
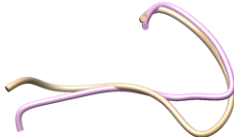


Standard Deviation (smaller is better)



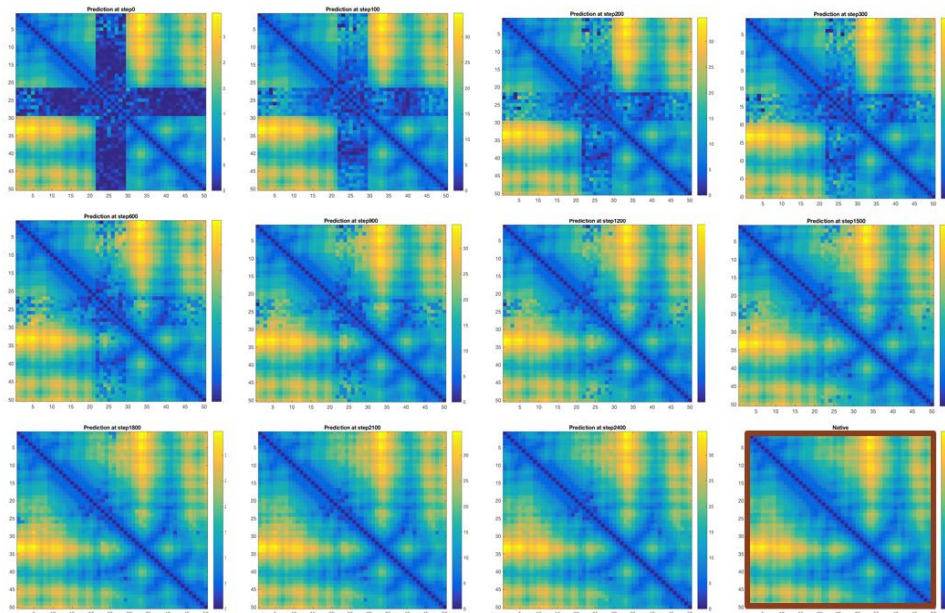
# Experimental Results

- Visualization of two structures *without GAN* and *with GAN* (yellow one is the native)

|                  | Without GAN  | With GAN   |
|------------------|--|--|
| 1CLC<br>(8 Res)  | <br>RMSD: 1.6 | <br>RMSD: 0.6 |
| 1BN8<br>(12 Res) | <br>RMSD: 3.0 | <br>RMSD: 1.7 |

# Experimental Results

- Visualization of *predictions* during the training process





# Conclusion

- **First successful application of GAN in bioinformatics**
- GAN makes the results more realistic and stable
- Distance map representation of protein allows us to treat it as an image
- Easy to apply deep neural networks on images
- Future work could be done to enable multiple loop modeling



# Thank you for listening

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