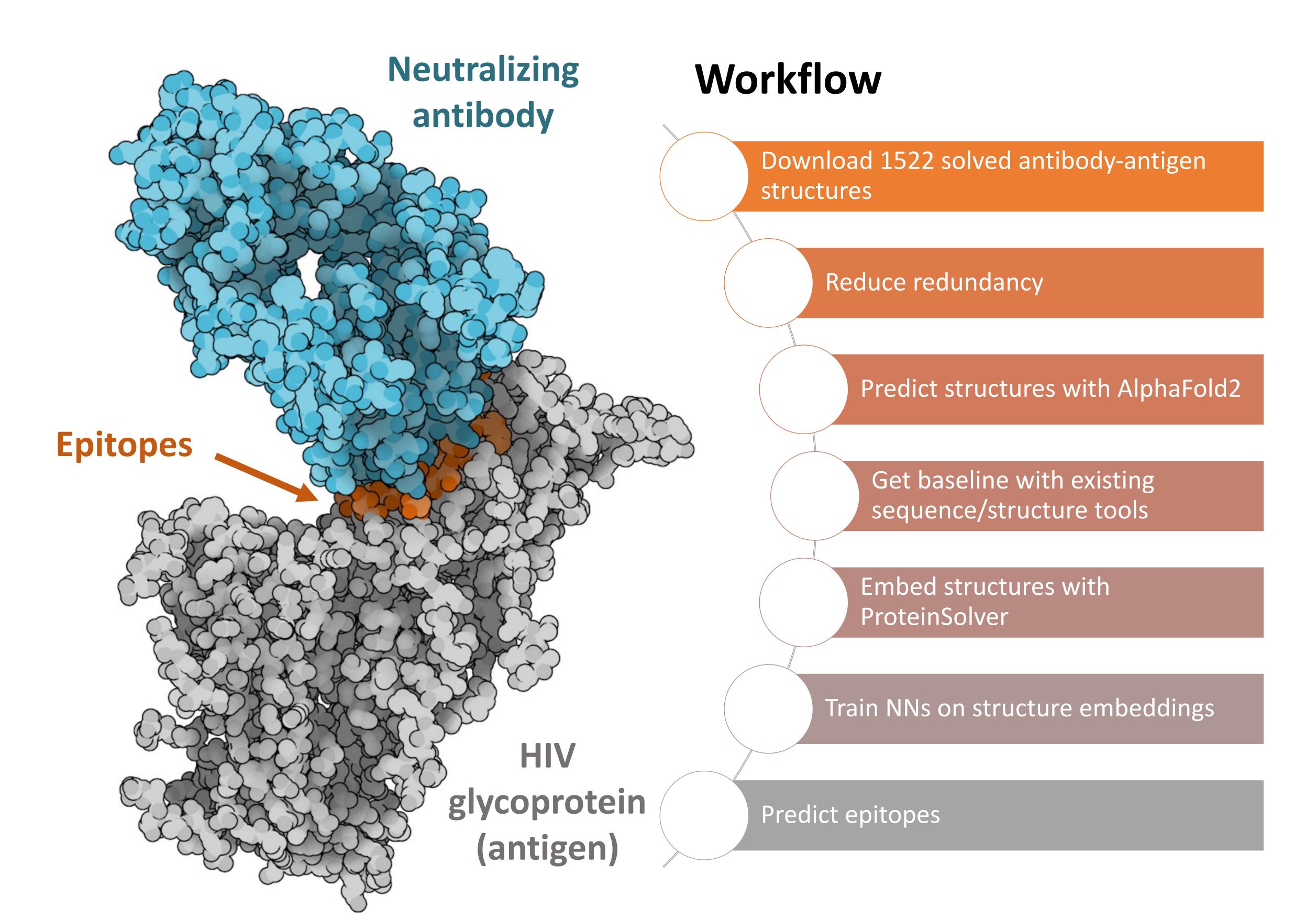


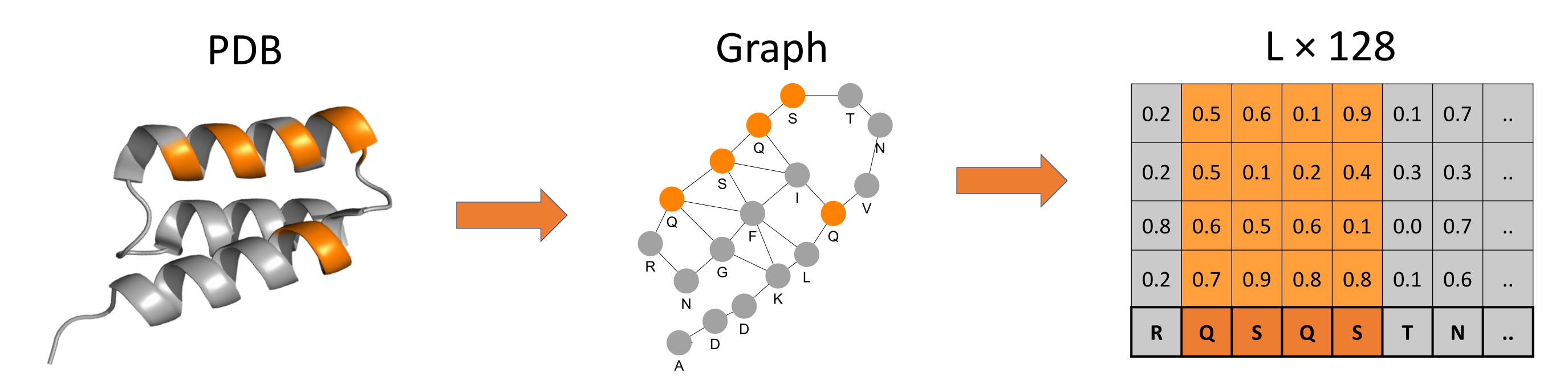
Prediction of epitopes using AlphaFold2 structures and graph neural networks

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Introduction

Epitopes are surface-exposed regions of a pathogenic molecule or antigen, which are targeted by the adaptive immune system by e.g. B-cell receptors. Binding is largely determined by the surface features of the target molecule. AlphaFold2 is a deep-learning protein folding model achieving near experimental quality prediction for many proteins. Furthermore, graph-based neural networks such as ProteinSolver allow structural representation of proteins suitable for tasks such as epitope prediction. We investigate improved epitope prediction using Alphafold2 modelled structures over sequence-only models.

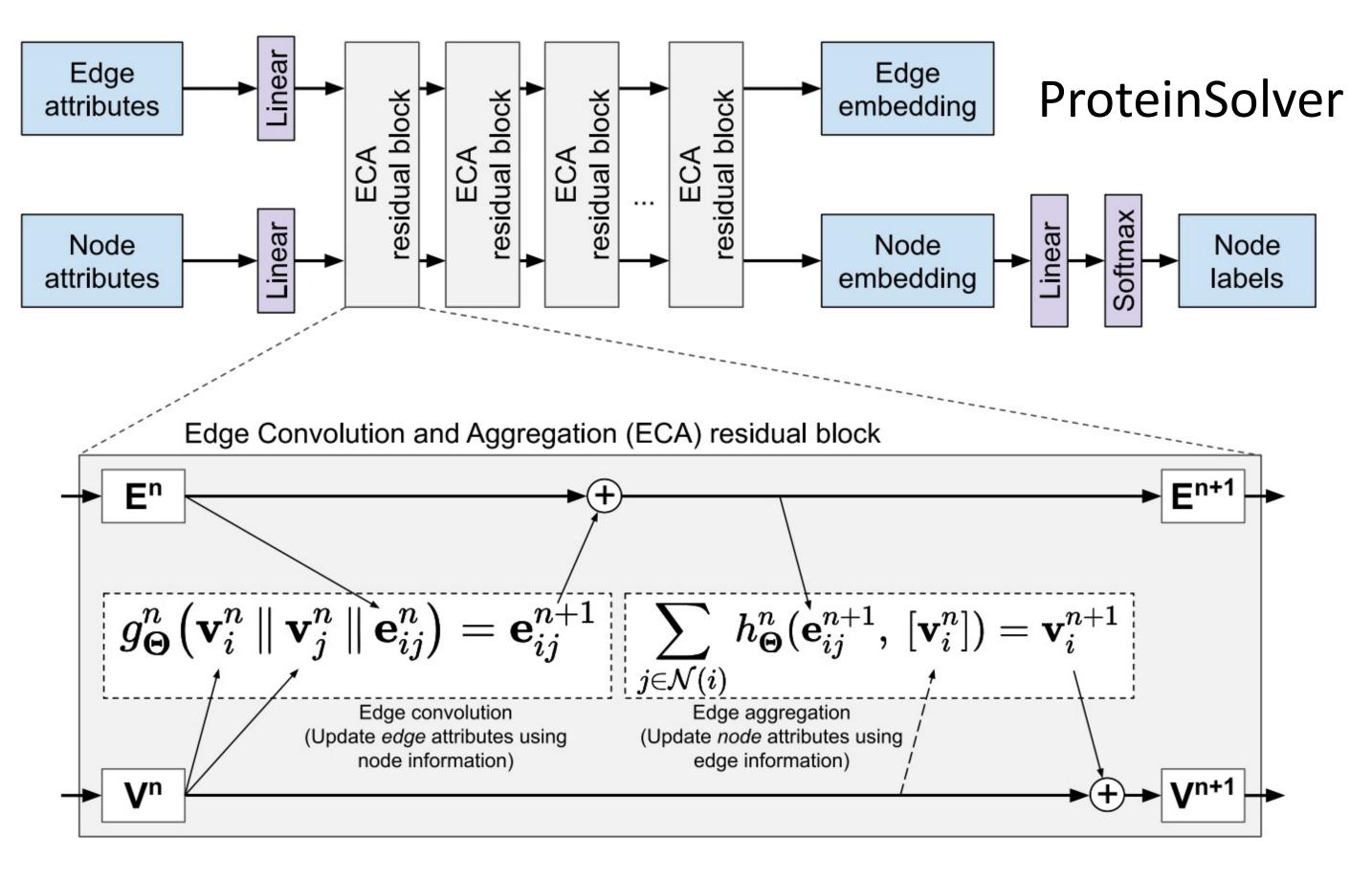




1. Model antigen structure with AlphaFold2 OR use experimentally solved structure

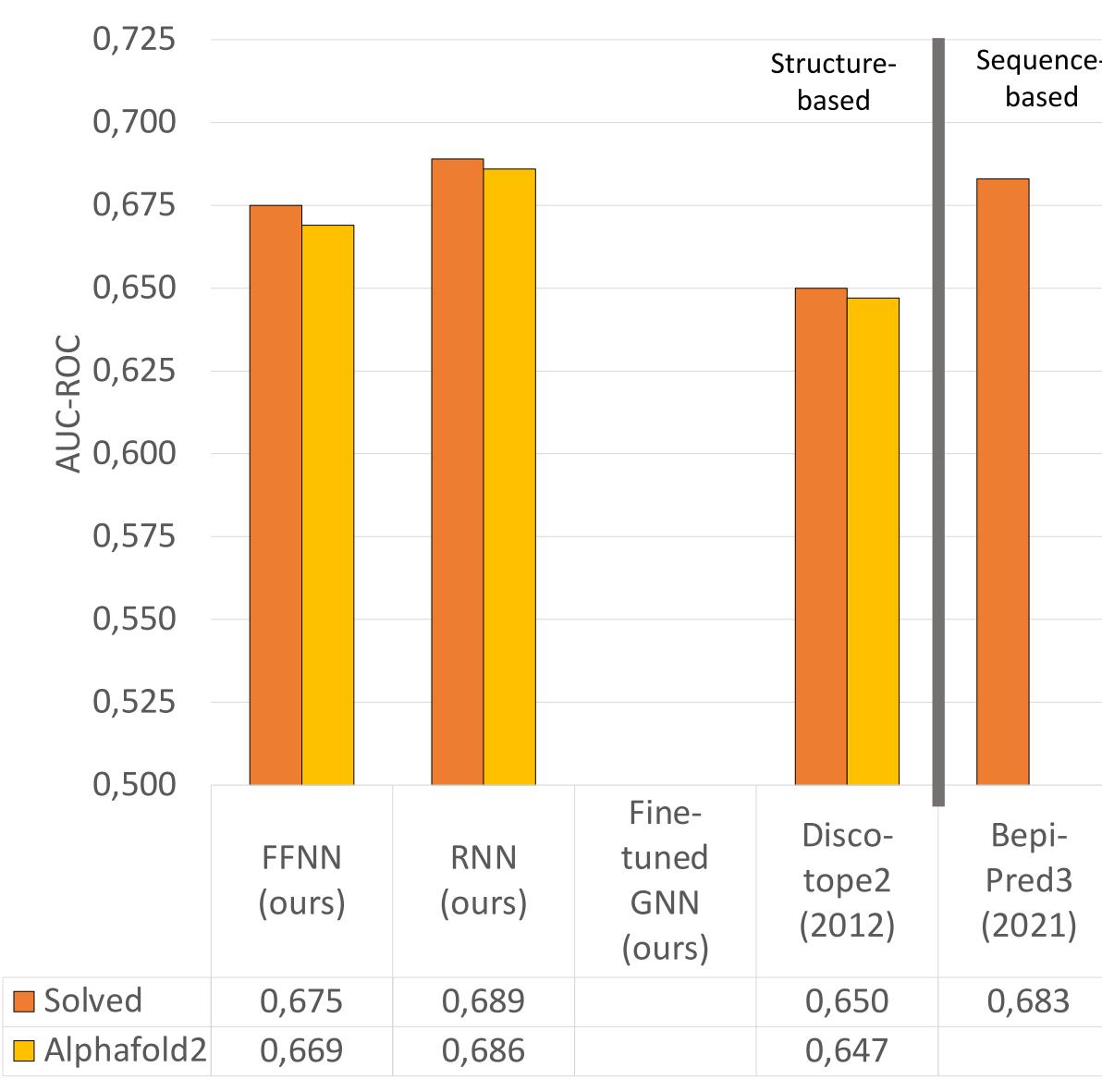
2. Embed structure in pretrained graph neural network (**ProteinSolver**) 3. Extract structure residue embeddings from nodes

Epitope prediction performance



Network architectures:

- FFNN Individual residues only (128)
- RNN Along sequence residues (L × 128)
- GNN Adapted from ProteinSolver (L × 128)
- Discotope2 Random forest model, surface accessibility
- BepiPred3 Sequence language model based on ESM



* Test set limited to PDBs deposited after AlphaFold2 training dataset collection date 30th April 2018

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