GuijunLab-QA GuijunLab-Human

## Evaluation of Protein Complex Quality using DeepUMQA-X Server

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# CONTENTS

— CONTENTS —

## 01 DeepUMQA-X 02 Components of DeepUMQA-X 03 Results



- Consensus EMA methods can show good performance when the quality of the model pool is good. Single-model EMA methods can show better performance than the consensus method when the quality of the model pool is poor.
- How to leverage the advantages of single-model methods and consensus methods to improve the performance of protein model quality assessment?

#### **DeepUMQA-X server Pipeline**



## Single-model EMA Methods

## Single-Model Method 1 for IDDT (GraphCPLMQA2)



Local assessment of complex model quality using enhanced network and Evoformer representation based on GraphCPLMQA (GuijunLab-Pathreader EMA)

GraphCPLMQA Liu et al., BIB 2024

#### **Single-Model Method 2 for TM-score**



Evaluation of protein complex quality based on multivariate representation using hierarchical networks (GuijunLab-Complex EMA)

### **Single-Model Methods of Dataset**



1,443,856 models

The construction process of monomer database and complex database

#### **Protein Features of Single-Model Methods**

#### **Geometry features**

- Triangle Position (USR improved version)
- Voxelization
- Distance and orientation
- Backbone angle and length
- Second structure

#### **Sequence features**

- Amino Acid Properties
- Relative position encoding

#### **Evolution features**

- Embedding of protein language models
- MSA Embedding of AlphaFold

#### **Statistical features**

- Rosetta energy
- Blosum62 score matrix



## **Consensus EMA methods**

### **Structural Alignment of Consensus Methods**

#### A naive way to obtain consensus information between multiple models: Structural alignment



How to quickly align large assemblies or massive complex model structures?

• Interface Sequence Align (At least 10 times faster)



### **QMODE3: Model Selection Pipeline**



- MassiveFold Sample Models (structural clustering)
- MassiveFold Sample Models-2 (MassiveFold Self-Assessment Selection)

## Results

### Case Study: T1201o - Homomer



### Case Study: T1201o - Homomer



#### **Case Study: H1213 - Heteromer**

Summary of submitted prediction models

Туре	Pearson	AUC	Top1Loss
SCORE	0.944	0.585	0.164
QSCORE	0.960	0.642	0.07
Local	0.789	0.812	-

#### Stoichiometry:A1B1C1D1E1 (1373AA)





#### **Case Study: H1213 - Heteromer**



## Case Study: H1233 - Antibody antigen

Summary of submitted prediction models

Туре	Pearson	AUC	Top1Loss
SCORE	0.972	0.919	0.01
QSCORE	0.952	0.877	0.02
Local	0.803	0.762	-

#### Stoichiometry: A2B2C2 (1316AA)



#### **Our Top1 Model** TM-score: 0.990 DockQ: 0.866





## Case Study: H1233 - Antibody antigen



### Case Study: H1227 - Large assembly

#### Summary of submitted prediction models

Туре	Pearson	AUC	Top1Loss
SCORE	0.886	0.5	0.15
QSCORE	0.807	0.5	0.08
Local	0.581	0.824	-





## **Case Study: H1227 - Large assembly**



#### **Case Study: H1204 - Nanobody Complex**



### Conclusions

#### Work

- DeepUMQA-X Server
- GraphCPLMQA2 (Improved version)
- Global scoring Method
- Interface Sequence Align (Lightweight Structure Align)

#### Assessment Challenges

- Large Assemblies
- Antibody-antigen complex
- Ensemble structures

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