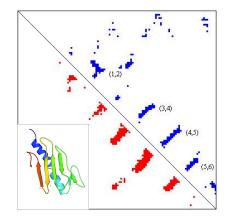
Protein Structure Modeling Guided by Deep Learning and Contact Prediction



The MULTICOM Group

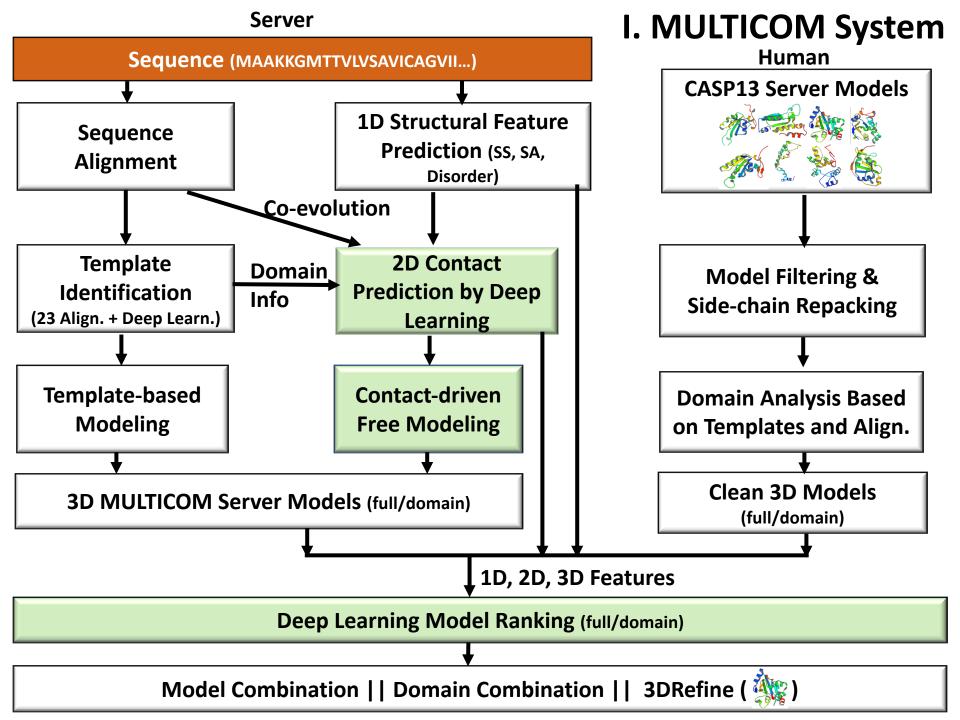


Jianlin Cheng

Department of Electrical Engineering and Computer Science University of Missouri - Columbia U.S.A. CASP13 Meeting, 2018

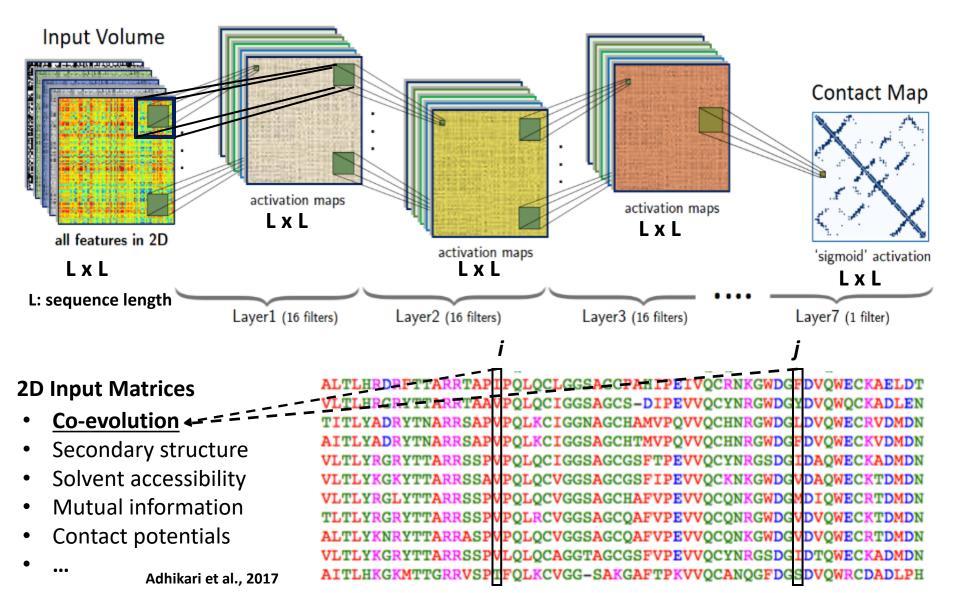
Outline

- Overview of MULTICOM system
- Three key new methods
- Analysis of examples
- •Summary



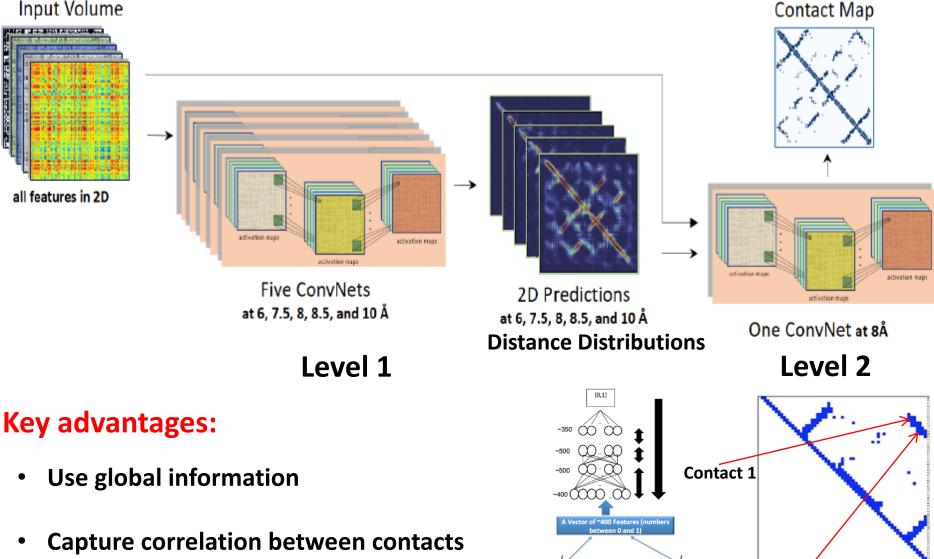
II. Three Key New Methods in CASP13

(A) 2D Convolutional Neural Network for Contact Prediction (DNCON2)



Two-Level Deep Convolutional Neural Networks

Input Volume



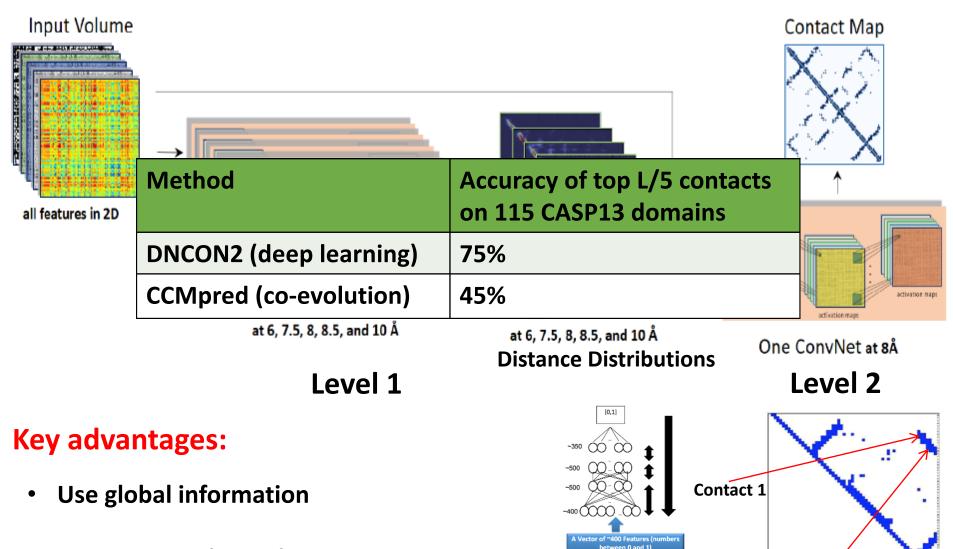
DEVYQYIVSQVKQYGIESRKYGDKAKYHLSQRW

Local Window

Contact 2

(high-level contact patterns / clusters)

Two-Level Deep Convolutional Neural Networks



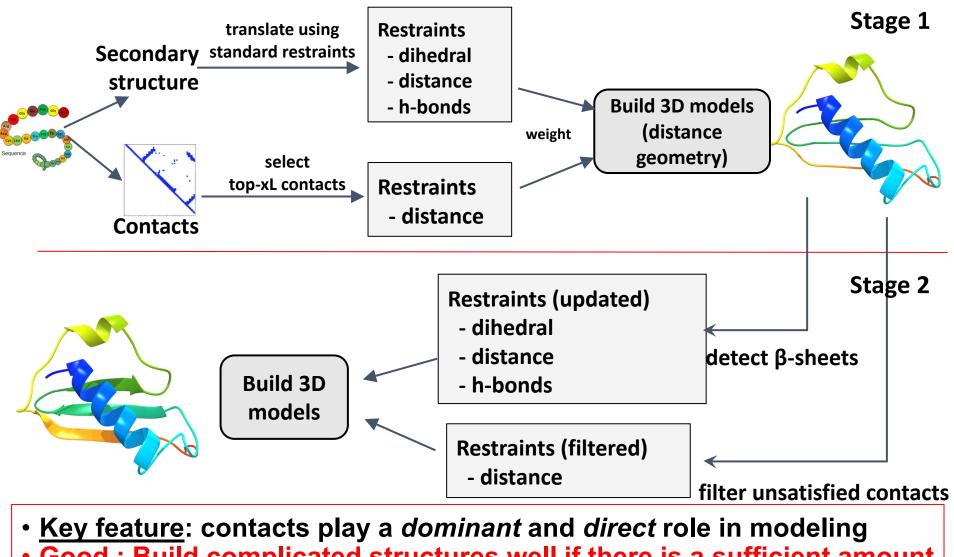
 Capture correlation between contacts (high-level contact patterns / clusters)



DEVYQYIVSQVKQYGIESRKYGDKAKYHLSQRW

Local Window

(B) Free Modeling by Translating Contact Distances into 3D Models (CONFOLD2)



• <u>Good</u> : Build complicated structures well if there is a <u>sufficient</u> amount of accurate distances; <u>Bad</u>: otherwise may fail Adhikari, Cheng, 2018

Free Modeling by Fragment Assembly with Contacts as Energy Terms

Rosetta + Contacts UniCon3D + Contacts FUSION + Contacts

Key feature: contacts are used as a part of an energy function to *indirectly* guide fragment assembly

- <u>Good</u>: use extra fragment information and energy, may work for small proteins with less complex topology
- <u>Bad</u>: fail if good models are not sampled, particularly for complicated structures

Comparison on T1000 – FM Domain (residues: 282-523)

DNCON2 (red) VS Native (blue) (L/5: 100%, L: 79%, 2L: 50%)

CONFOLD (red) VS Native (L/5: 67%, L: 65%, 2L: 55%)

250 250 250 △ input △ input native native 300 300 300 350 350 350 400 400 400 450 450 450 500 200 500 Top-2L Top-2L Top-2L 250 350 400 450 500 300 450 500 250 300 350 400 300 250 350 400 **Purple: model** Green: native TM-score: 0.80 TM-score: 0.33 GDT-TS-score: 0.23 GDT-TS-score: 0.64 Top L/5 contacts on native structure

Rosetta-Con (red) VS Native (L/5: 20%, L: 18%, 2L: 17%)

input

native

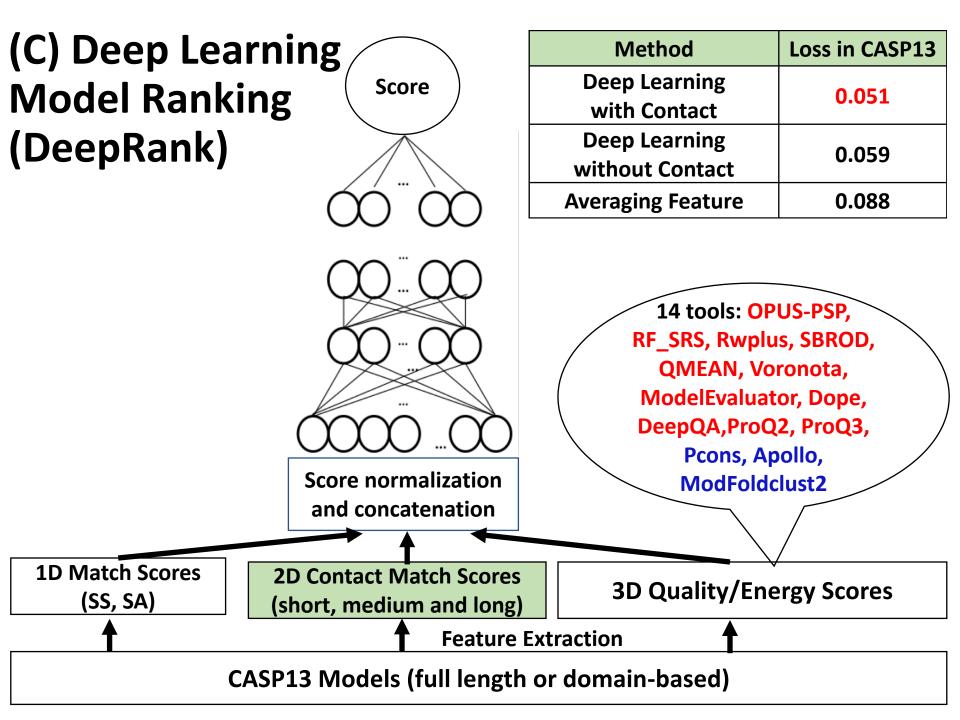
п

450

500

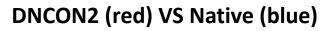
Red: model

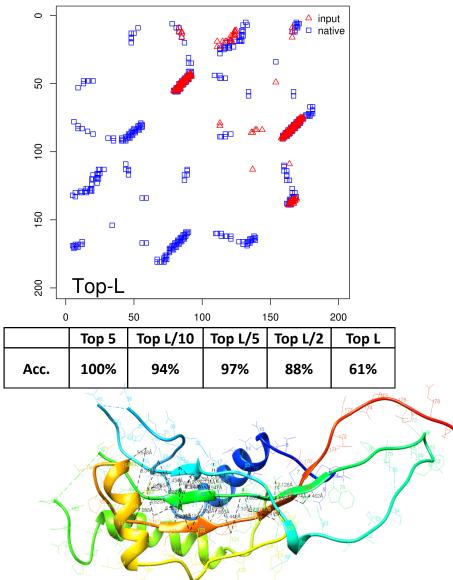
Green: native



III. Analysis of Examples (Success and Failure)

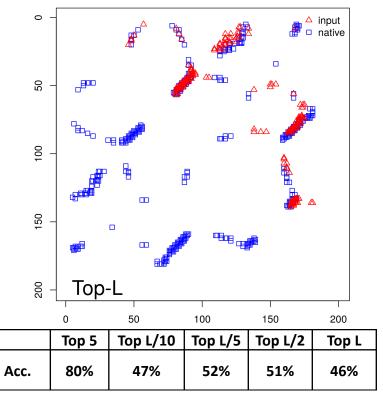
(1) Success of Building Models for T1021s3-D1 (FM) by CONFOLD

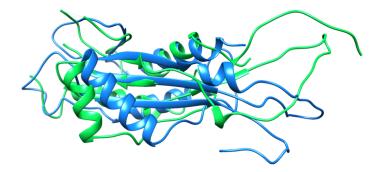




Top L/5 long-range contacts on native structure

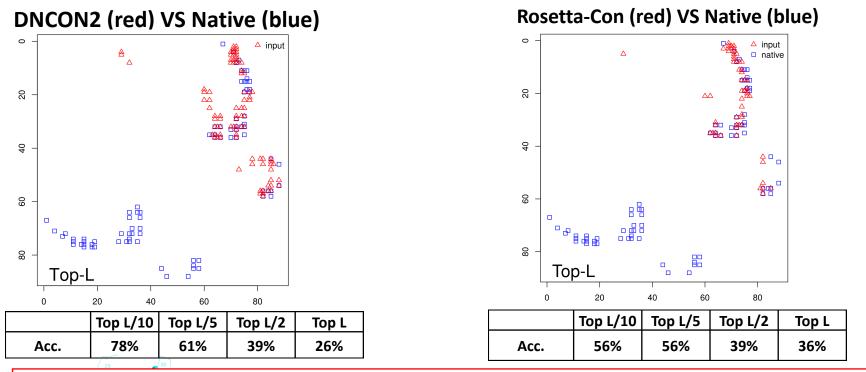
CONFOLD (red) VS Native (blue)





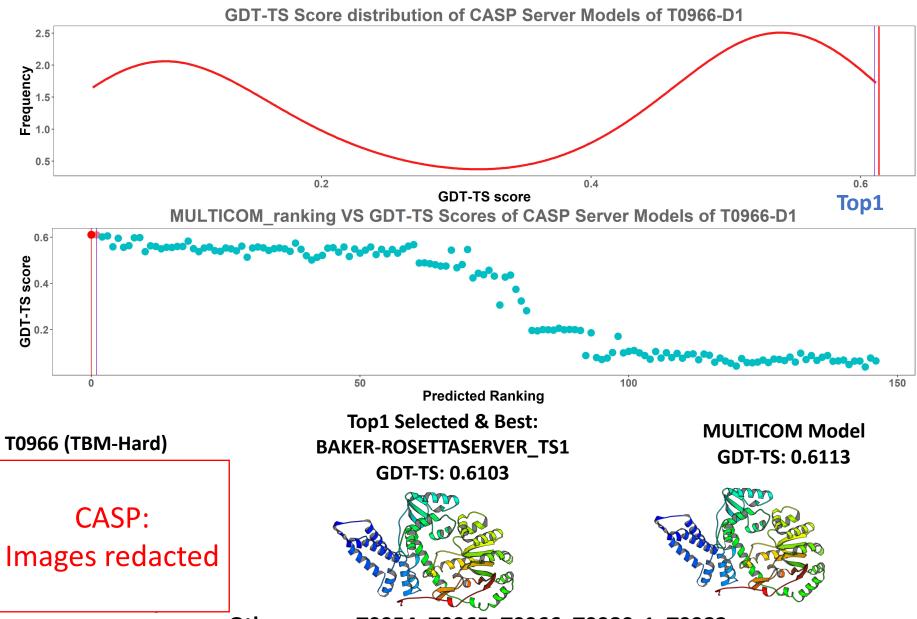
Blue: predicted; Green: native TM-score: 0.50 GDT-TS-score: 0.41

(2) Success of Building Models from Contacts with Rosetta When Failing to Identify Templates for T1019s2 (TBM)



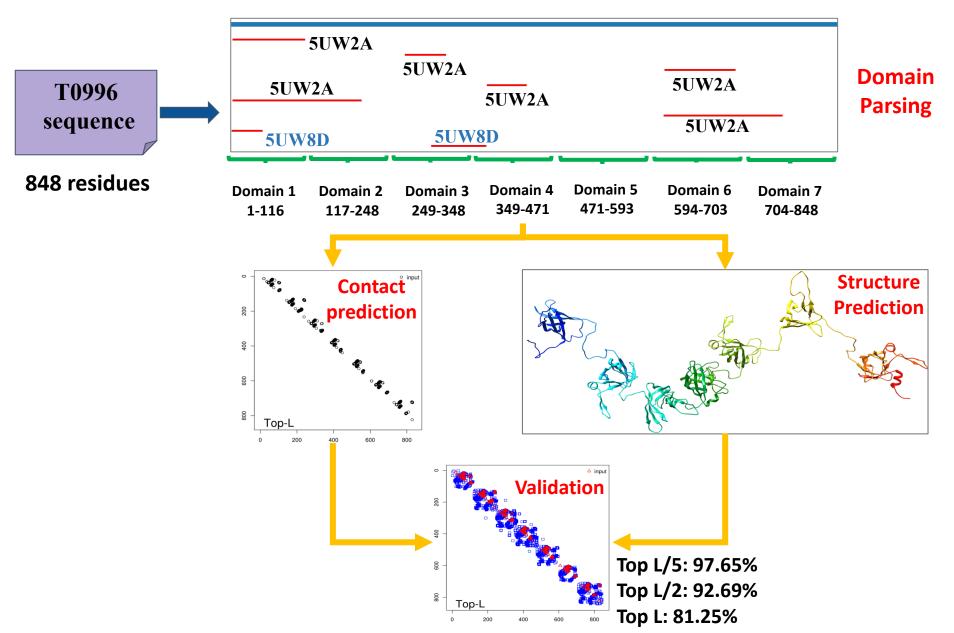
CASP: Images of protein structures redacted

(3) Success of Model Ranking and Combination with Deep Learning

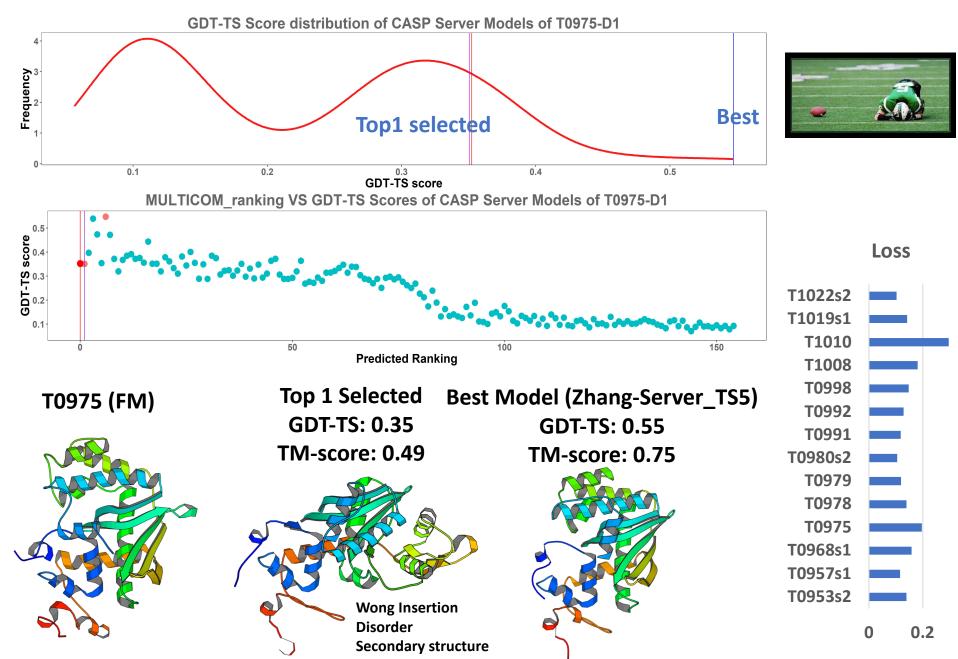


Other cases: T0954, T0965, T0966, T0980s1, T0982

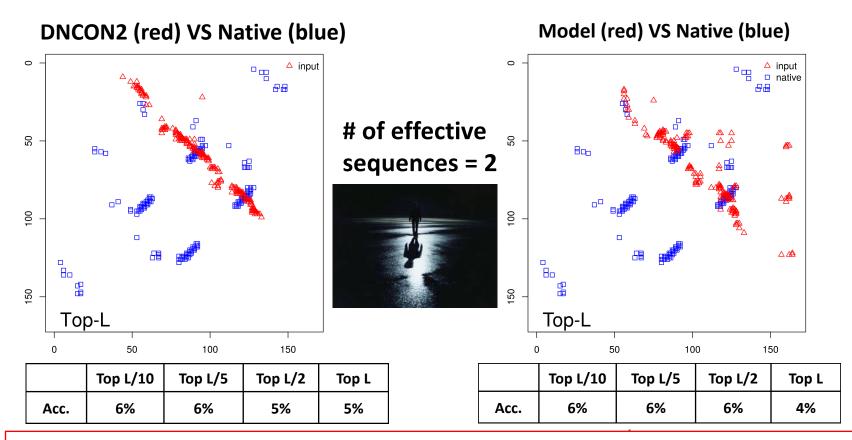
(4) Success of Domain Parsing, Template Identification, Domain-Based Model Ranking with Deep Learning for T0996 (TBM)



(1) Failure of Ranking Models (Loss >=10)



(2) Failure of predicting / using contacts (T0998 FM)



CASP: Images of protein structures redacted

Top L/5 medium-range contacts on native structure TN

TM-score: 0.21 GDT-TS-score: 0.15

IV. Summary

- Contact/distance prediction is the light in the dark world of free modeling.
- Contact prediction is valuable for ranking models and templates.
- Deep learning holds the key of protein structure prediction.
- Contact/distance-based free modeling and fragmentbased free modeling are complementary.
- There are significant challenges in model ranking and contact/distance-based modeling.

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