PLMSearch: Protein language model powers accurate and fast sequence search for remote homology

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Overview

- Introduction to Homologous
 Protein Search
- Background Knowledge / Related Works
- Introduction to PLMSearch
- Methodology
- Results / Experiments
- Future Work
- Questions / Discussion

Homologous Protein Search

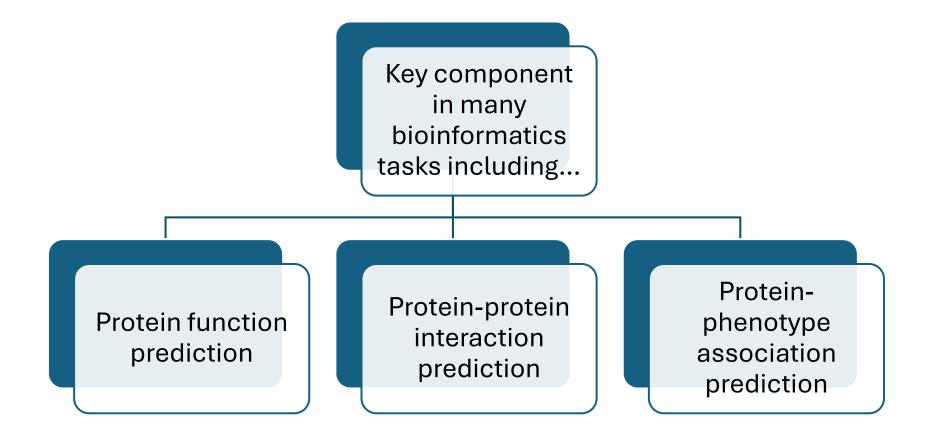
The goal of homologous protein search is to associate a query protein with homologous proteins

OHO

Homologous Protein Search

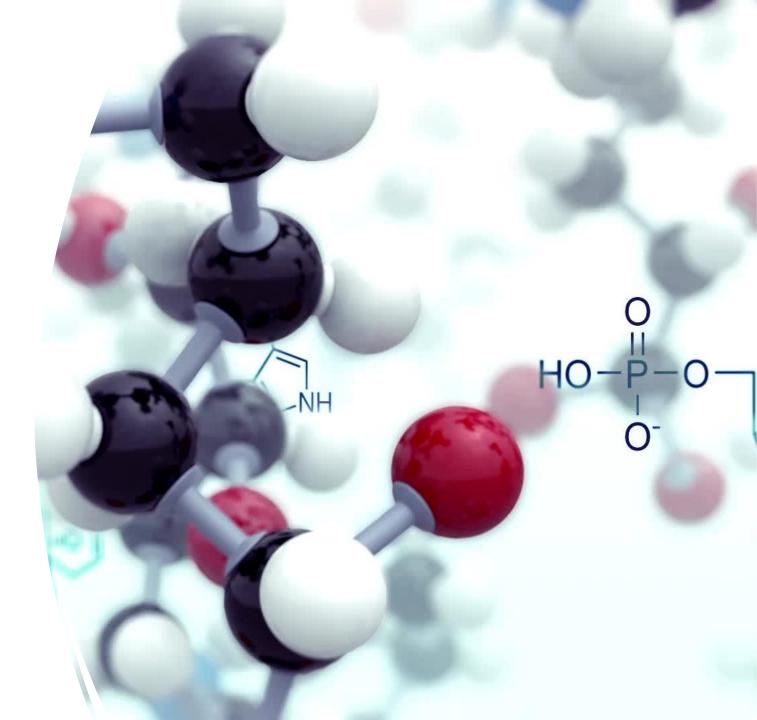
Two homologous proteins share evolutionary origins

Homologous Protein Search



Key Challenges for Homologous Protein Search

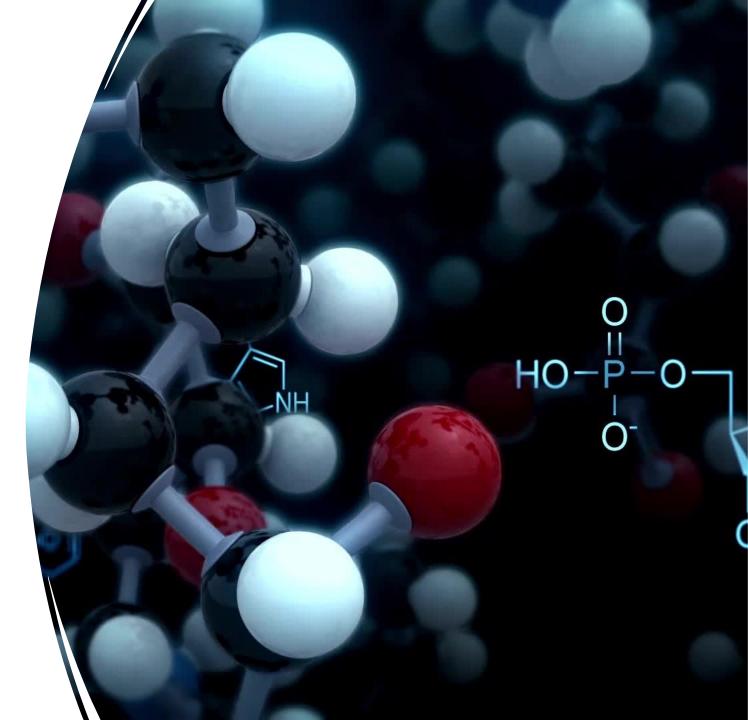
- 1) Computational Complexity
- 2) Noisy Datasets
- 3) Low Sequence Similarity



Homologous Protein Search

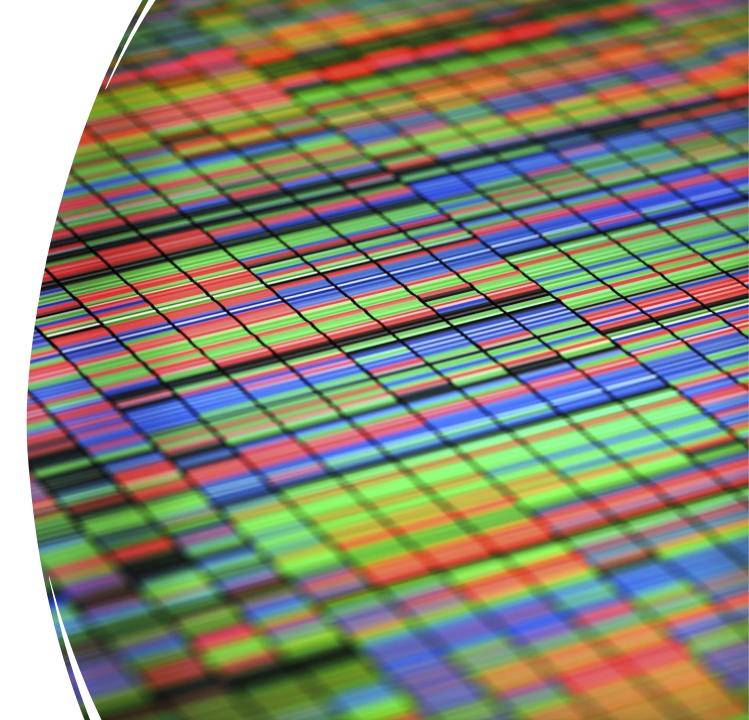
Homologous protein search can be roughly divided into two approaches or components

- 1) sequence search
- 2) structure search



Sequence Search Approach

Goal: Compare amino acid sequence of query to targets in database



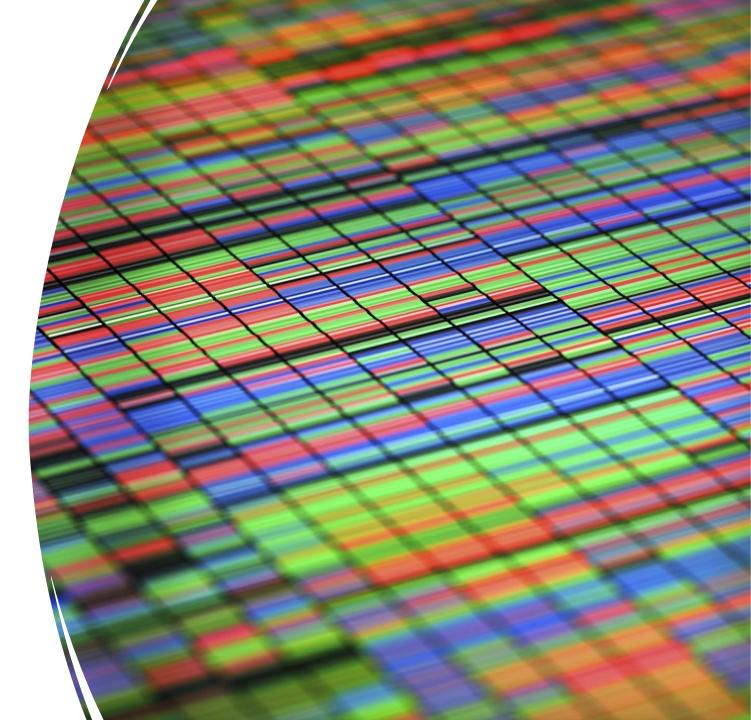
Sequence Search Approach

Strengths

- 1) Low cost
- 2) Data abundance
- 3) Well established

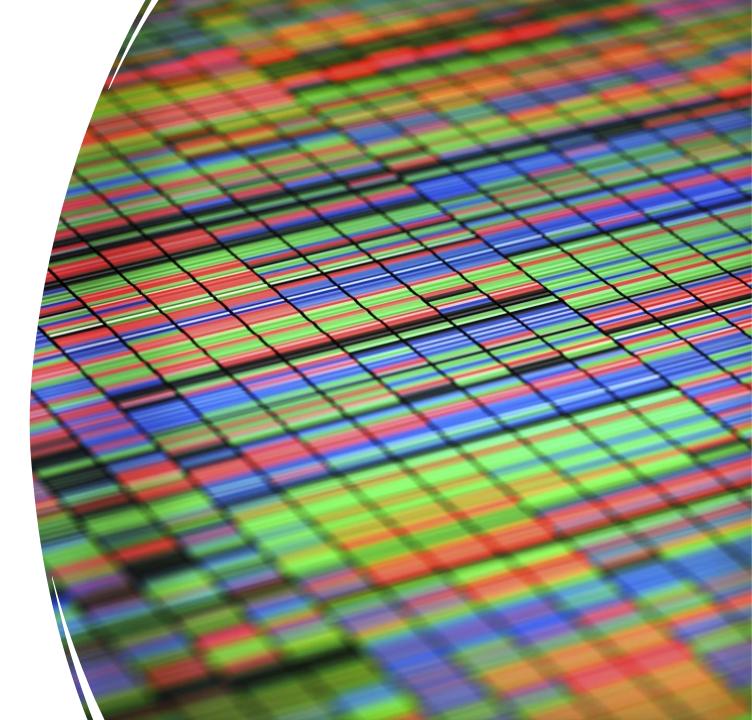
Challenges

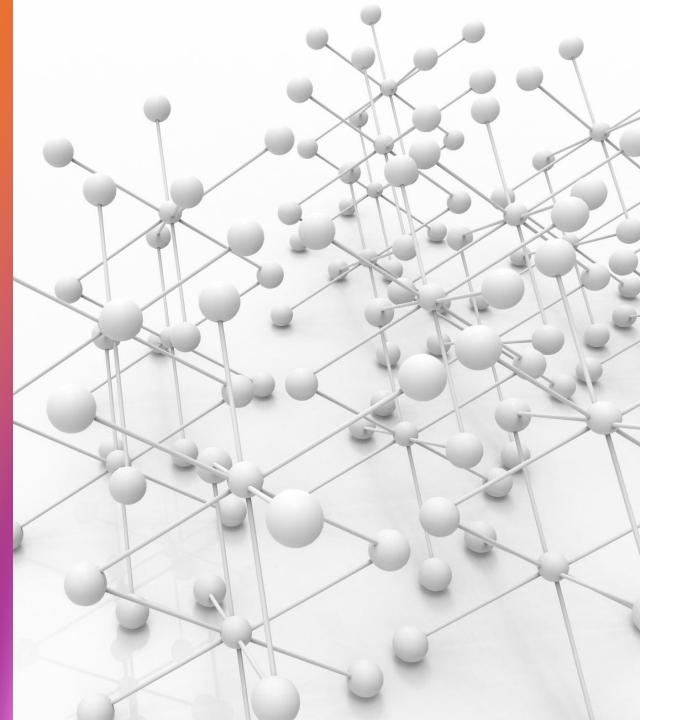
- 1) Sequence divergence
- 2) Poor remote performance



Sequence Search Approach

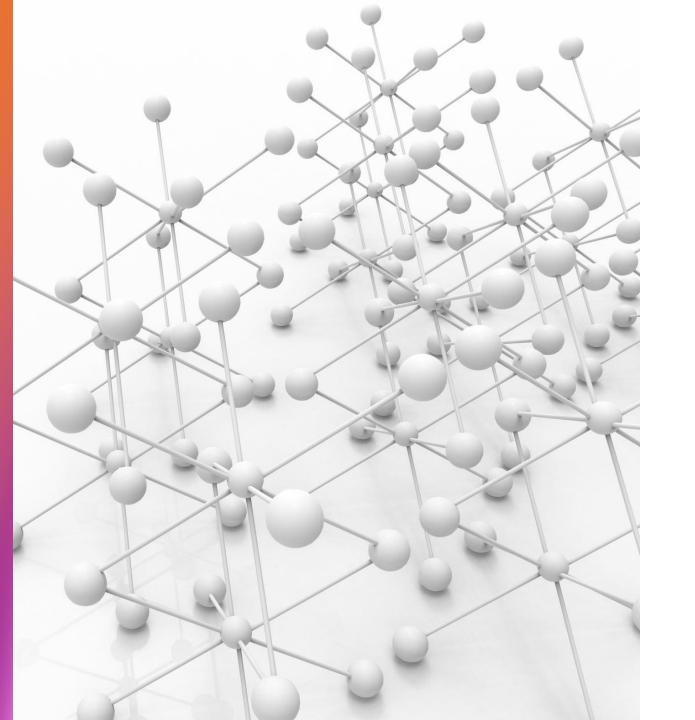
- MMseqs2
- BLASTp
- Diamond





Structure Search Approach

• **Goal:** Perform the comparison of the query and target proteins leveraging 3D structure of the proteins



Structure Search Approach

Strengths

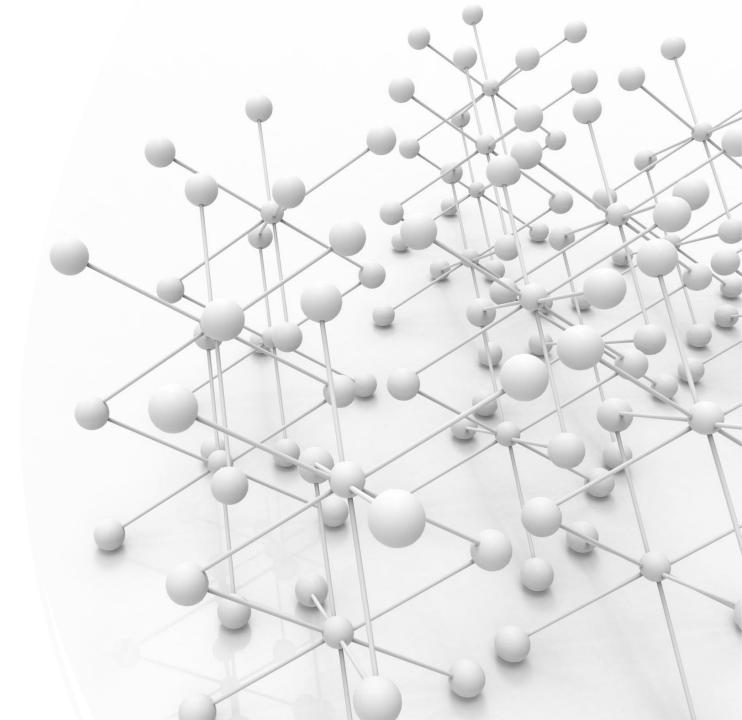
1) Better remote performance

Challenges

- 1) High cost
- 2) Data scarcity

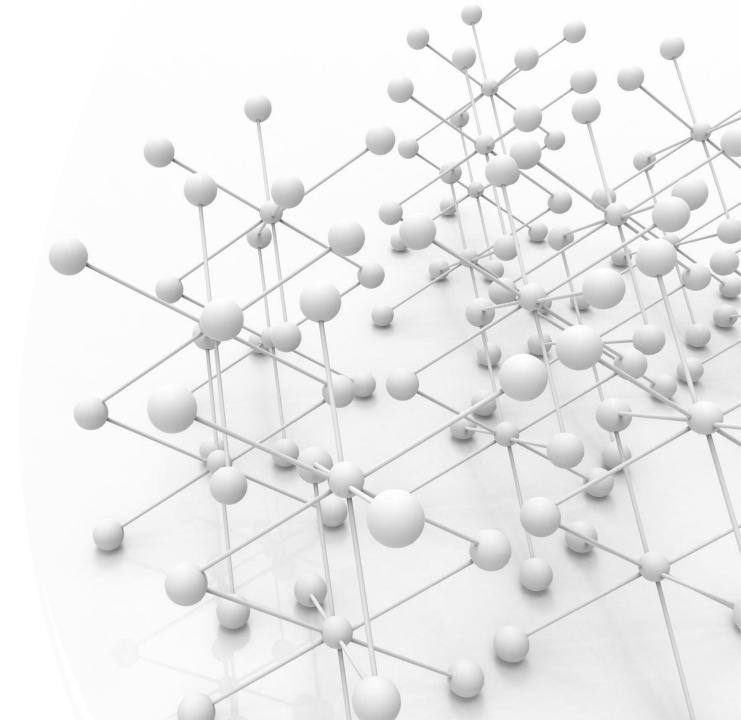
Structure Search Approach

Protein structure search methods can be further divided into 3 main approaches



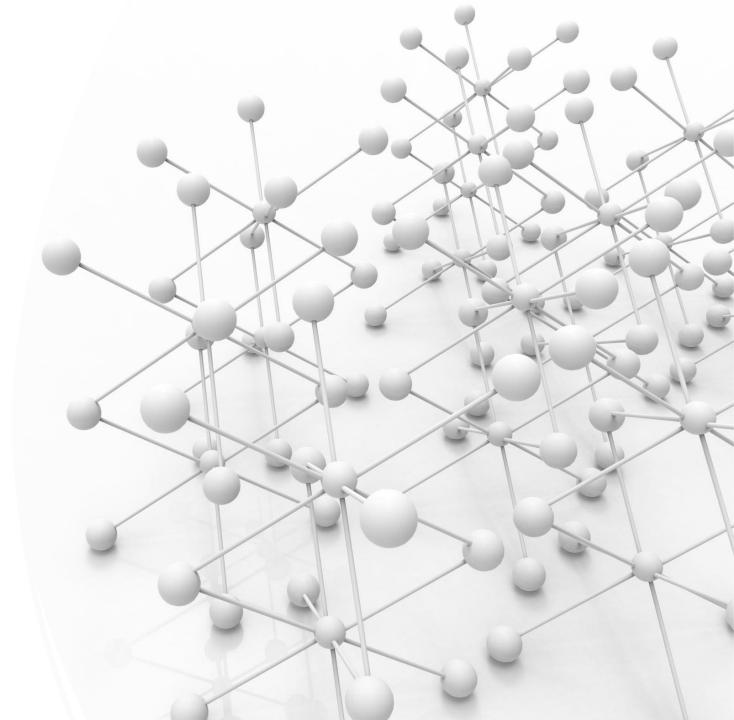
Contact/Distance Map-Based

- Map_align
- EigenTHREADER
- DiscoVER



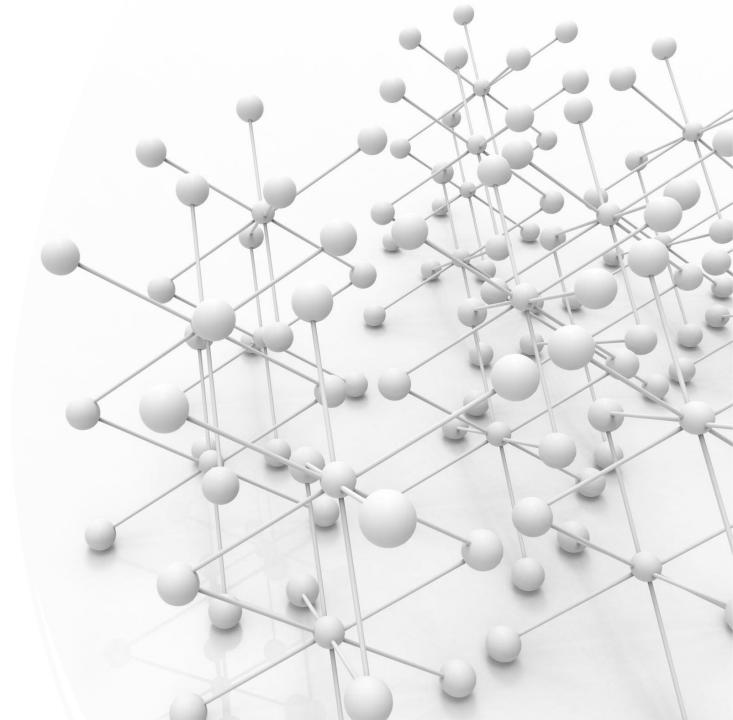
Structural Alphabet-Based

- 3D-BLAST-SW
- CLE-SW
- Foldseek
- Foldseek-TM



Structural Alignment-Based

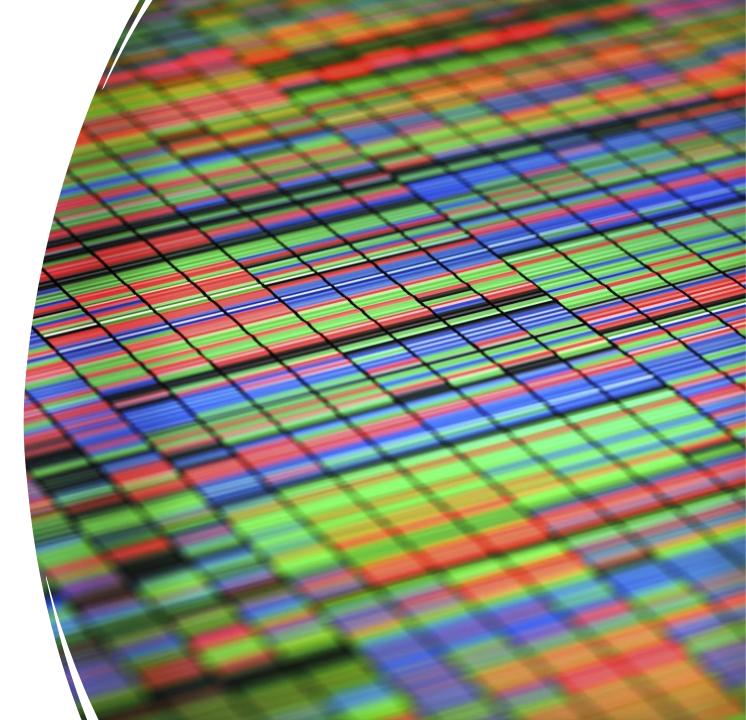
- CE
- Dali
- TM-align



An Old Fix: HMM

Hidden Markov Models (HMMs)

- HMMER
- HHsearch
- HHblits



HEDTNTHLWIVRHAMEIMANNKDVVKPGEVEQLKQWQSDLEQGIYDADHAN SYWHWKGTNPEDWLHGTAVAAKKDYPDIVNDTTKAWFVKAAVSNSYAAKWR/ A090Y8D7 A0A090Y8D7 BACMY Phospholipase C domain pro INEGVSSHLWIVNELIDIMSONT **LEWRTDLEKGIYSADYO** SYWNWKSANPEEWV **VAAKA** NFLKAAVSQDSADKWR/ Q87 J8RQ87 E Pho OS=Bacillus cereus DEGVSTHL RNELEKGVYSADYE ATV/ SYWNWK: VN LDAAISQDAADKWR/ 073JVAD C OS=Bacillu Pho ase AEGKNS AIDIMARD **WRTDLEDGIYTADYER** AA WO PGTVM SYWN SAEYANMWR OK1F nge phos AARA *QWQQGLYDADFK* NESEST 1VDRG7 IA C SPPSGP ILQURNIDWRFVERQHVDYRDCWE BRO C3GB BAC se C OS=Bacillus thurir FHNLKNAFEQGLYDADHL DVNONTHLWIARHAID WIHYAAANAKVHAKNIIRP SAFEEYVTEIQHNFNDLQNTM AOAOWUN9 AOAOAOWUN9 9BA holipase C OS=Bacill Ph INVNESTHLWLAQDAINRLARNOD NAAAFFPEYKTSFEQGLYDADYLI LPYYYSYDSGDWIHFAALIAKI IREEIFEYGSLSDFKIRLNNKWK R6C7|Q6R6C7 LISIV Phospheripase C OS=Listeria ivanov

The New Wave: Protein Language Models

(PLM)

Goal: Learn contextual representations of proteins by treating sequences as language

DTNTHLWIVRHAMEIMANNKDVVKPGEVEQLKQWQSDLE HWKGTNPEDWLHGTAVAAKKDYPDIVNDTTKAWFVKAAV 0Y8D7 A0A090Y8D7 BACMY Phospholipase C GVSSHLWIVNELIDIMSONT LEWRTDLE INWKSANPEEW **AVAAKA** NFLKAAV 7 J8RQ87_E Pho OS=Bacil GVSTHL RNELE INWK: VN LDAAI 3JVAD Pho base C BAG GKNSE AIDIMARD TEWRTDLE AA WOU PGTVN INI (1R 50 AARA SEST 1VDRG D A DWO AK SGP RILO RNIDWRFVERO 10 C3GB BAC se C OS=Bacil RD. NQNTHLWIARHAID **SEFHNLKNAFE** EEYVTEIQHNFNDLQNTM WIHYAAANAK WUN9 A0A0A0WUN9 9BA nolipase C <P /NESTHLWLAQDAINRLARNOD NAAAFFPEYKTSFE YYYYSYDSGDWIHFAALIAKI **1IREEIFEYGSLSDF** 7|Q6R6C7_LISIV Phospherepase C OS=Liste

Protein Language Models

- ESMs
- ProtTrans
- ProtENN
- CATHe
- DEDAL
- DeepBLAST
- pLM-BLAST



PLMSearch

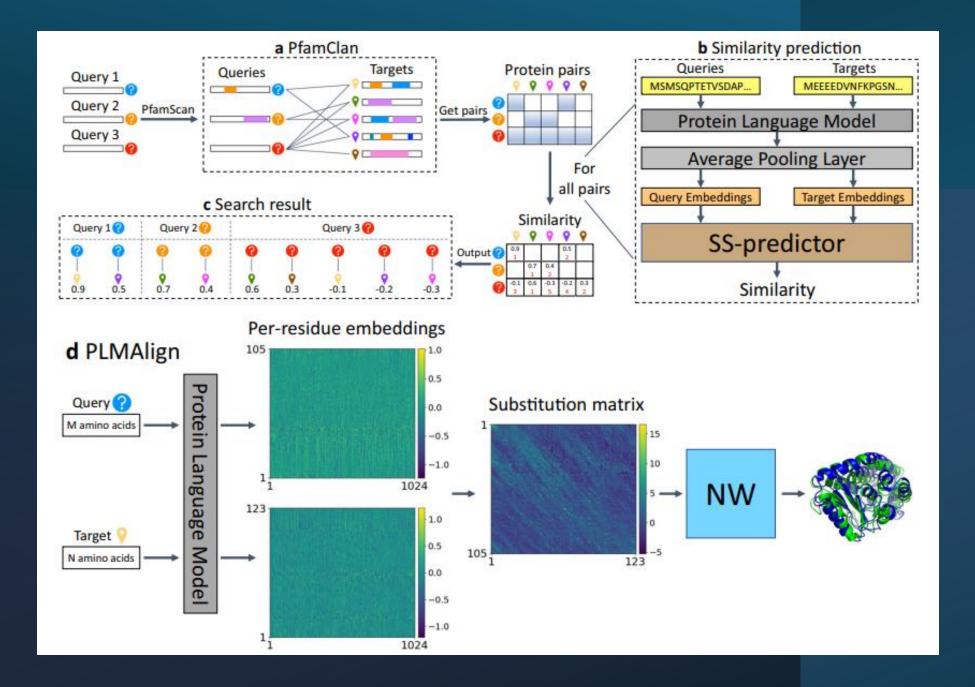
PLMSearch is designed to improve the **sensitivity** of sequence-based protein searches while maintaining their **efficiency** and **universality**.

PLMSearch High Level

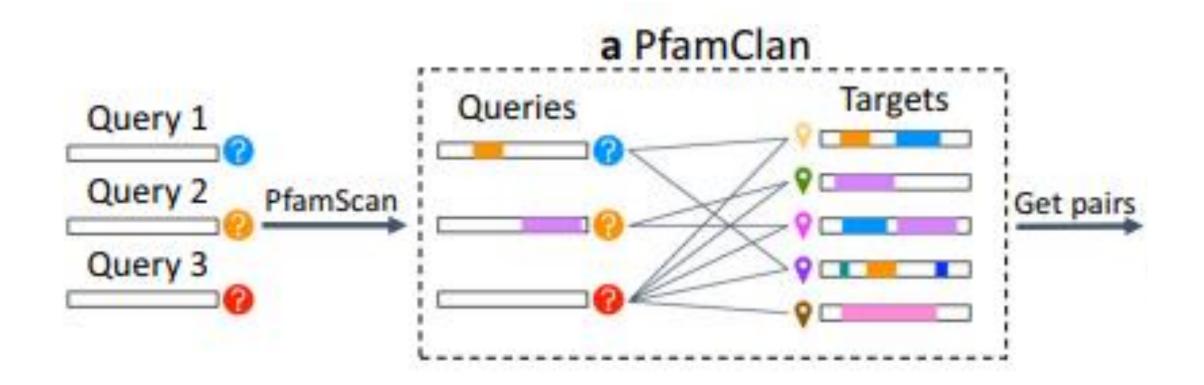


1) PFAMCLAN FILTERING 2) SS-PREDICTOR

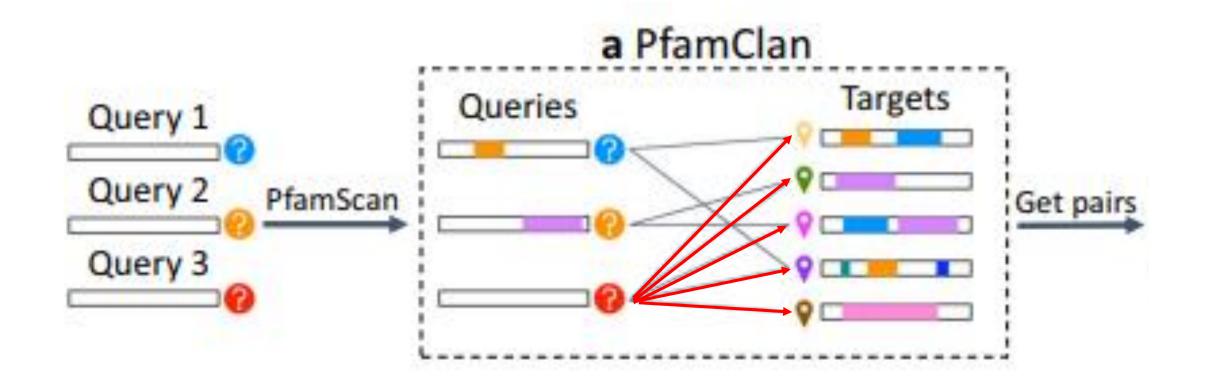
3) RANKING / PLMALIGN



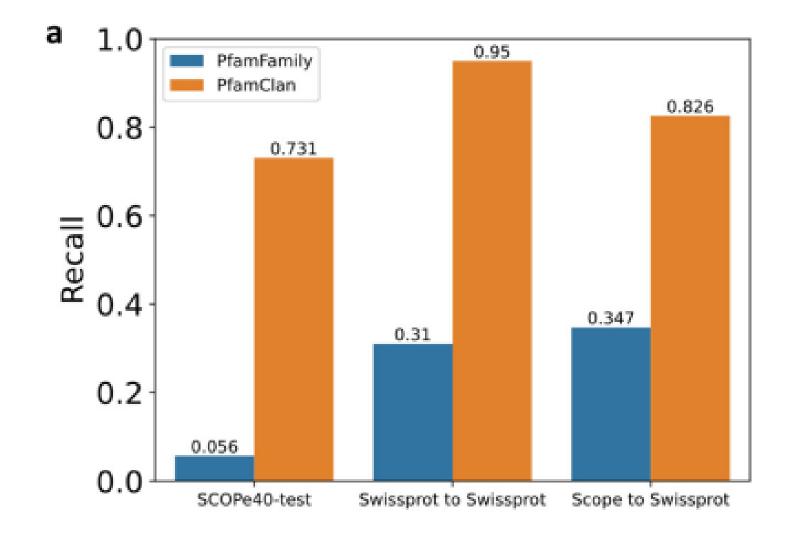
PfamClan

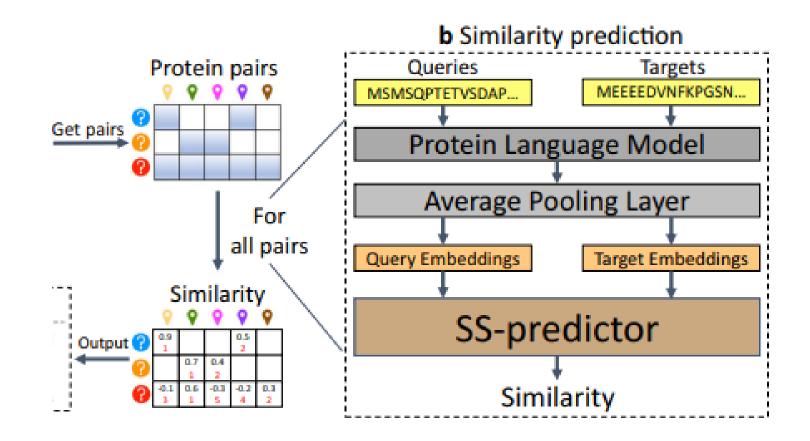


PfamClan

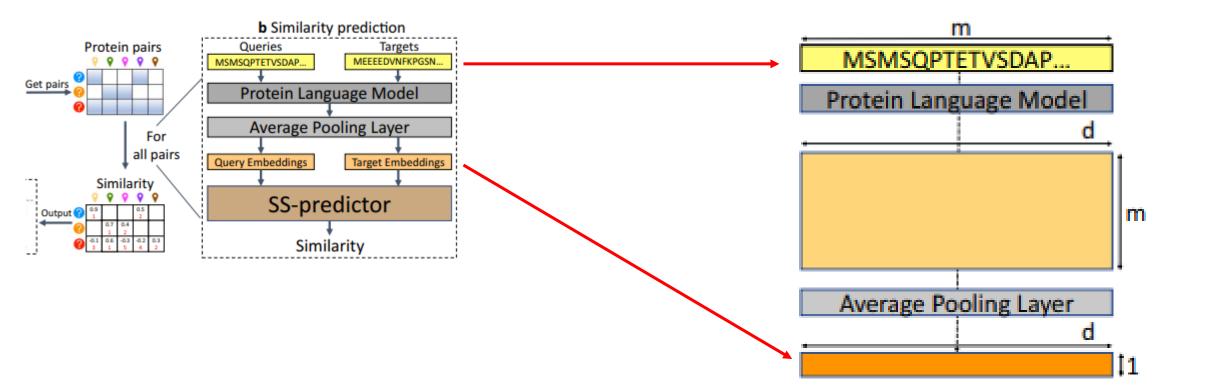


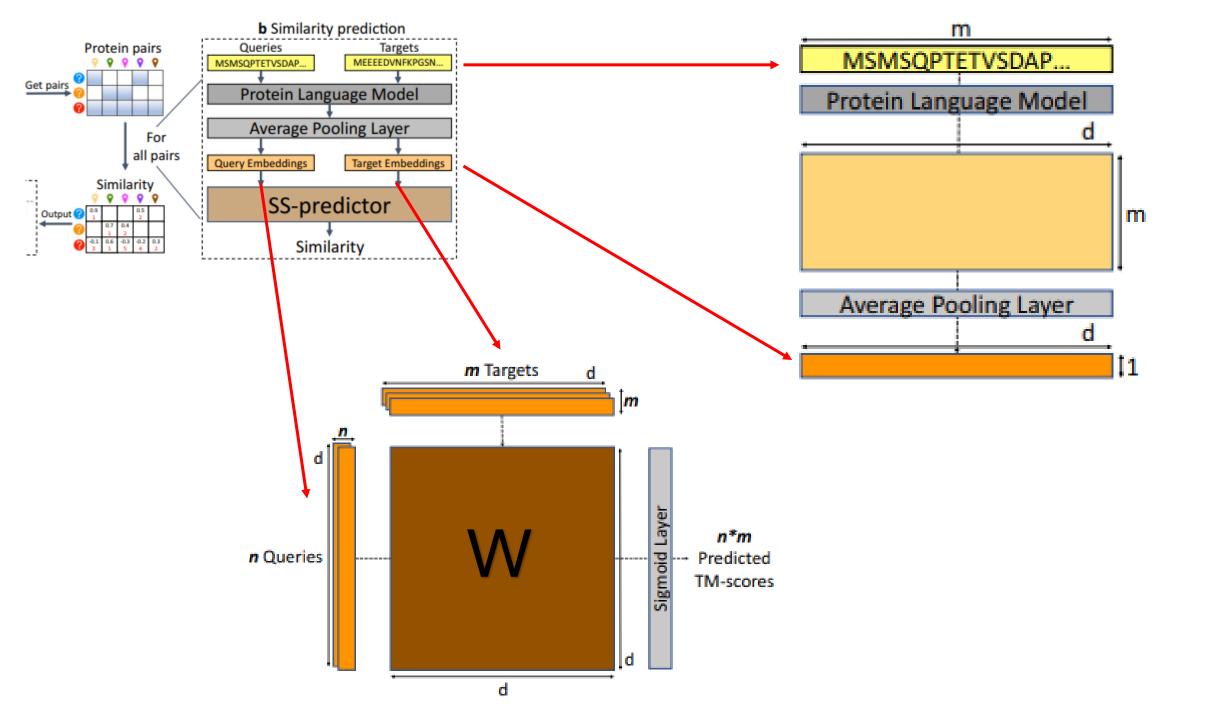
Pfam Clans vs Pfam Families

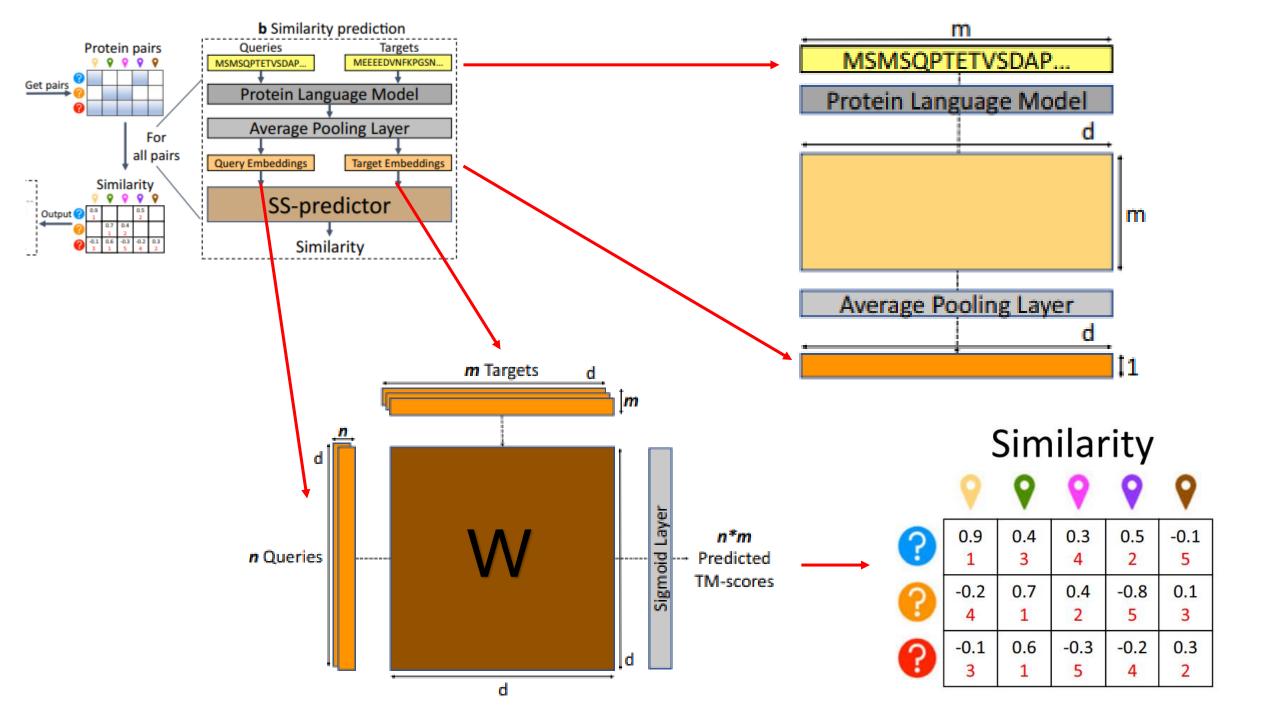




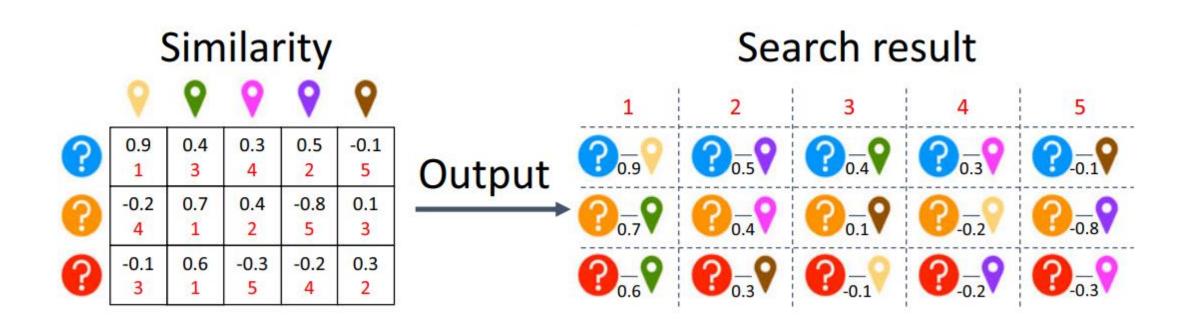
Structural Similarity Predictor (SS-predictor)





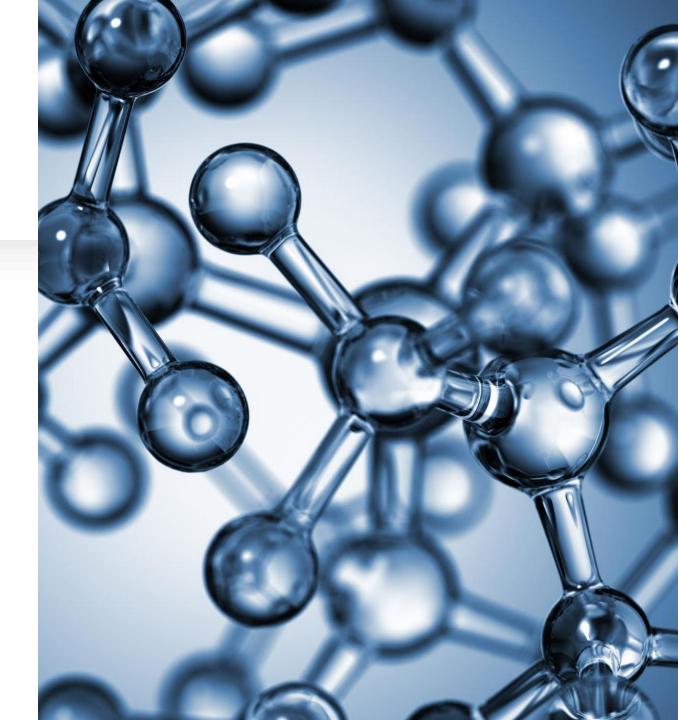


SS-predictor: Final Output

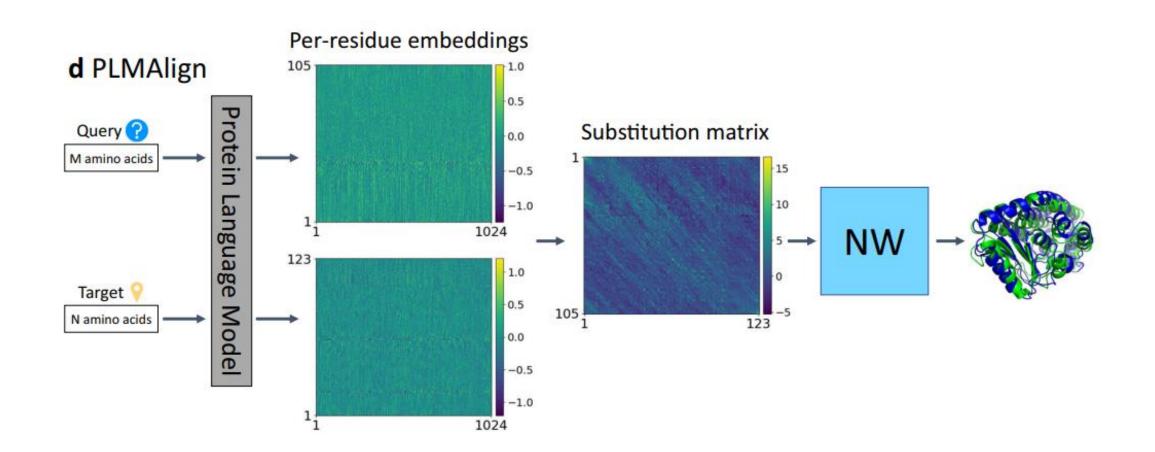


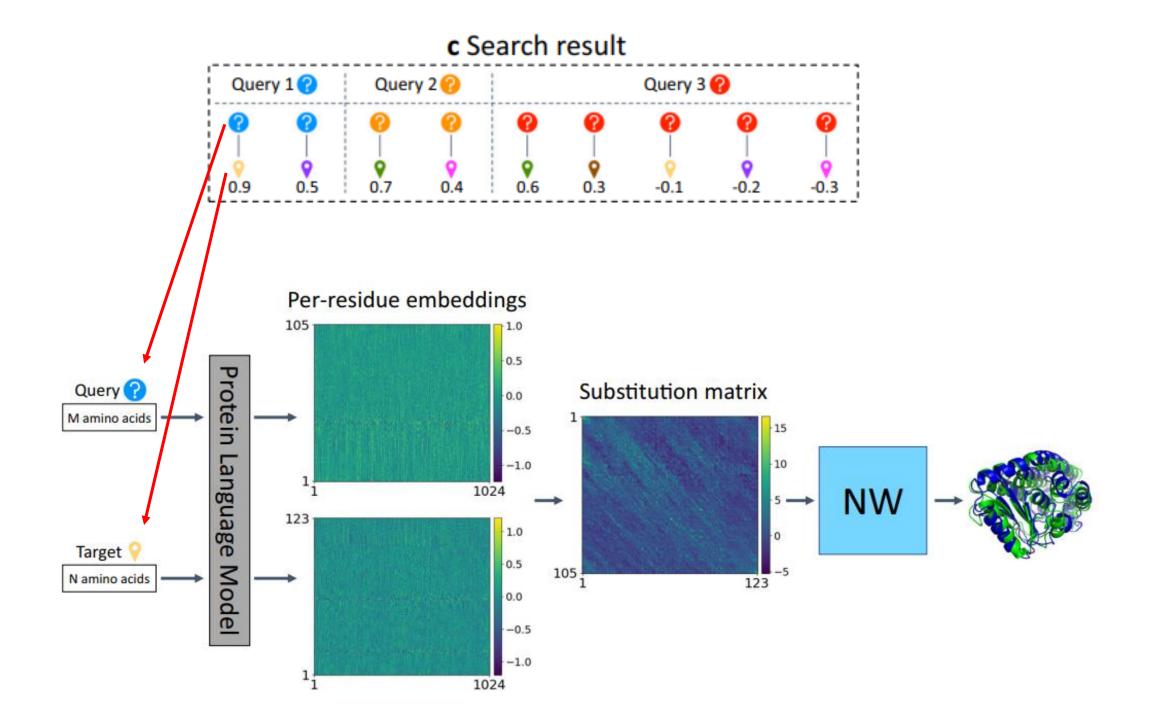
SS-predictor: Key Advantages

- 1) Efficiency: Avoids the computational cost of structure-based alignment for large datasets
- 2) Scalability: Handles the increasing size of protein clusters effectively through sequence-based predictions
- 3) **Precision:** Combines TM-score and COS similarity to handle both low and high sequence identity cases
- 4) Global Similarity Focus: Optimized for detecting global structural similarity, making it suitable for applications requiring a complete fold detection



PLMAlign





Evaluation/Results

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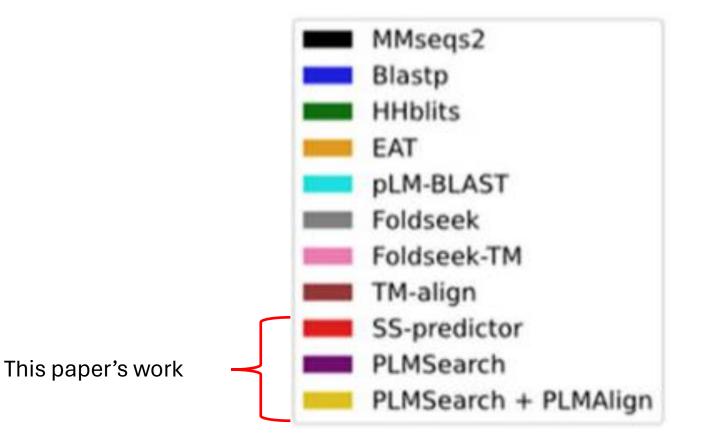
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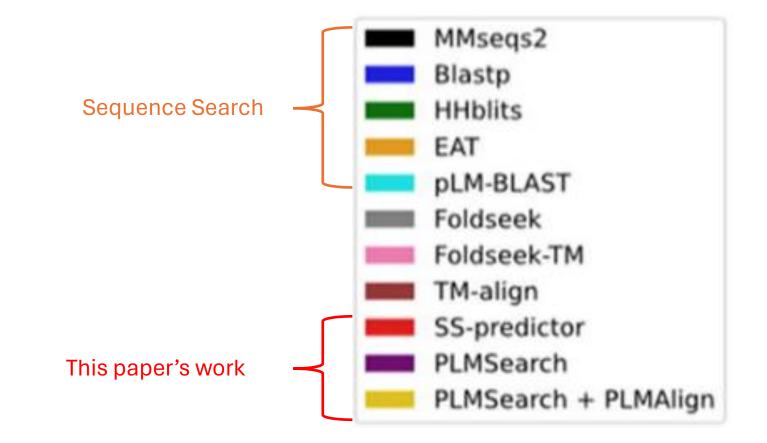
A

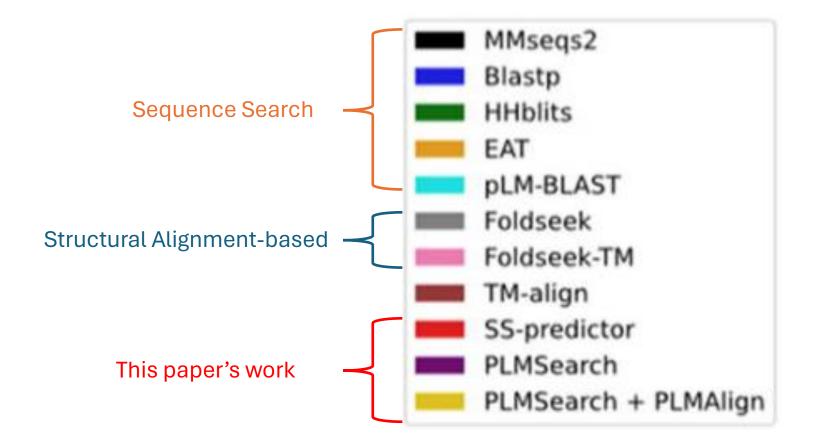
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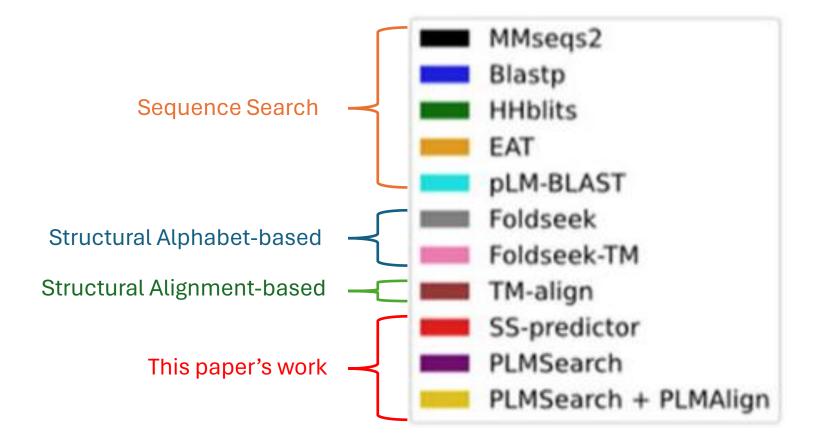
Dx)

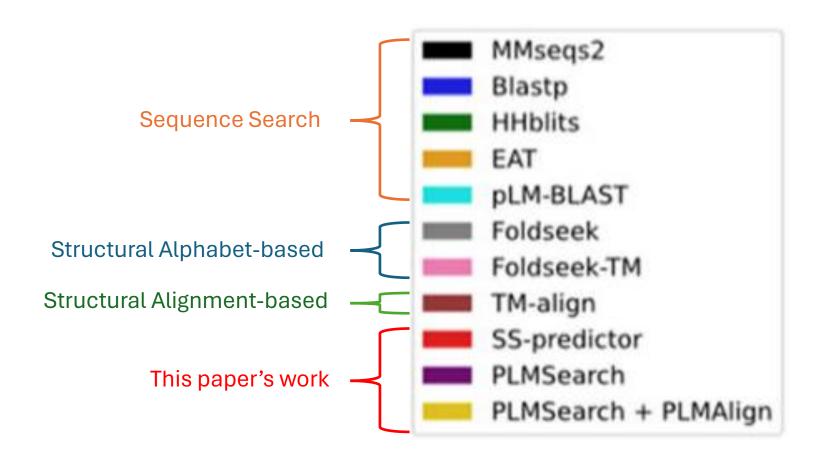
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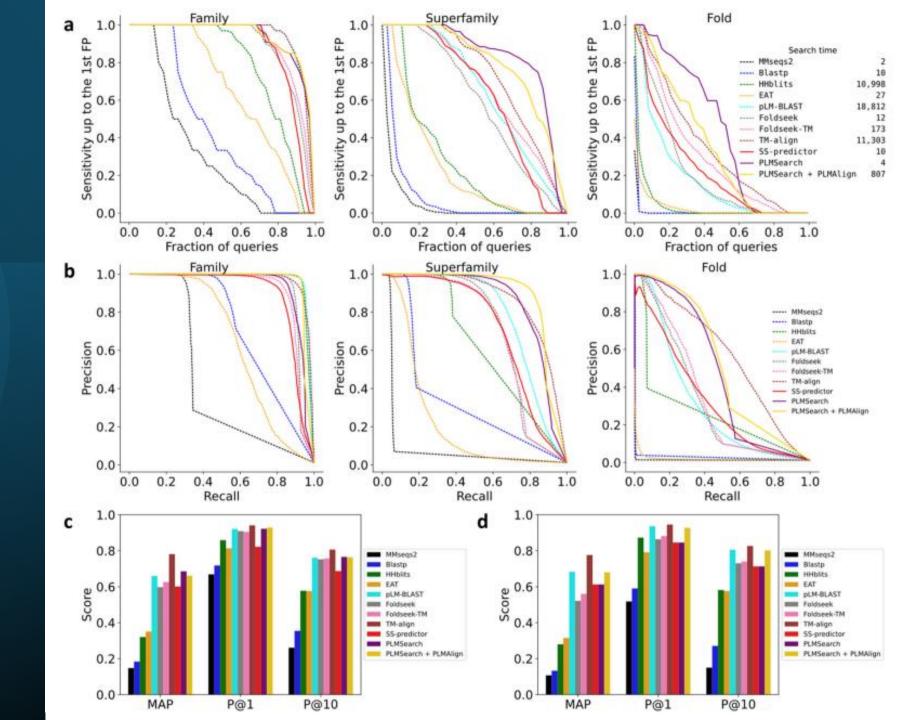




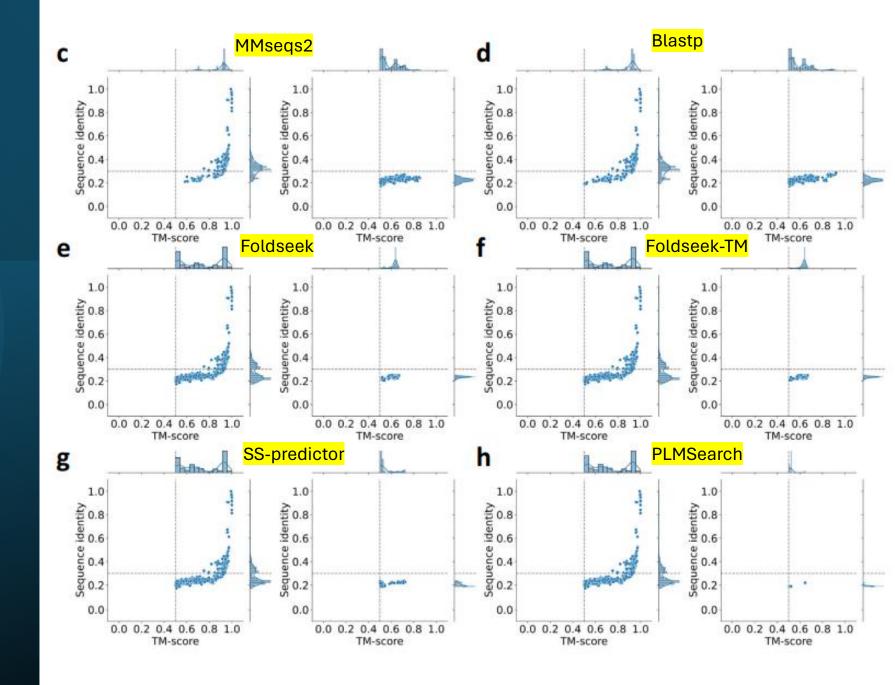


- SCOPe40-test dataset (2207 proteins)
- All-versus-all search test is performed
- 4,870,849 total query-target pairs

All-vs-All Search Test on SCOPe40-test



All-vs-All Search Test on SCOPe40-test





Results Summary

- PLMSearch is comparable to S.O.T.A. structural search methods
- PLMSearch overcame low sensitivity cap on sequence methods
- Improvements are concentrated in remote homology pairs
- PLMSearch is one of the fastest search methods
- Residue embedding-based alignment methods are limited by size of target dataset

Future Works

Efficient Embedding Storage





Thank You!

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Questions?