

# **PLASMA: Fast and Interpretable Protein Substructure Alignment via Optimal Transport**

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# Local structural motifs are the key to understanding protein function

Active sites, binding pockets, and structural motifs link structure to function. Structural conservation is 3–10× stronger than sequence conservation.

With AlphaFold DB growing rapidly, we need scalable tools to uncover conserved motifs across the protein universe.

# Existing methods have critical limitations



## Template-based methods

Cannot discover novel structural similarities

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## Global alignment (TM-Align, Foldseek)

Computationally demanding, difficult to scale

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## Embedding-based methods

Compress residue info → poor local interpretability

# And therefore here's our solution...



Cannot discover novel structural motifs

**Optimal transport alignment — no predefined templates needed**



Computationally demanding, difficult to scale

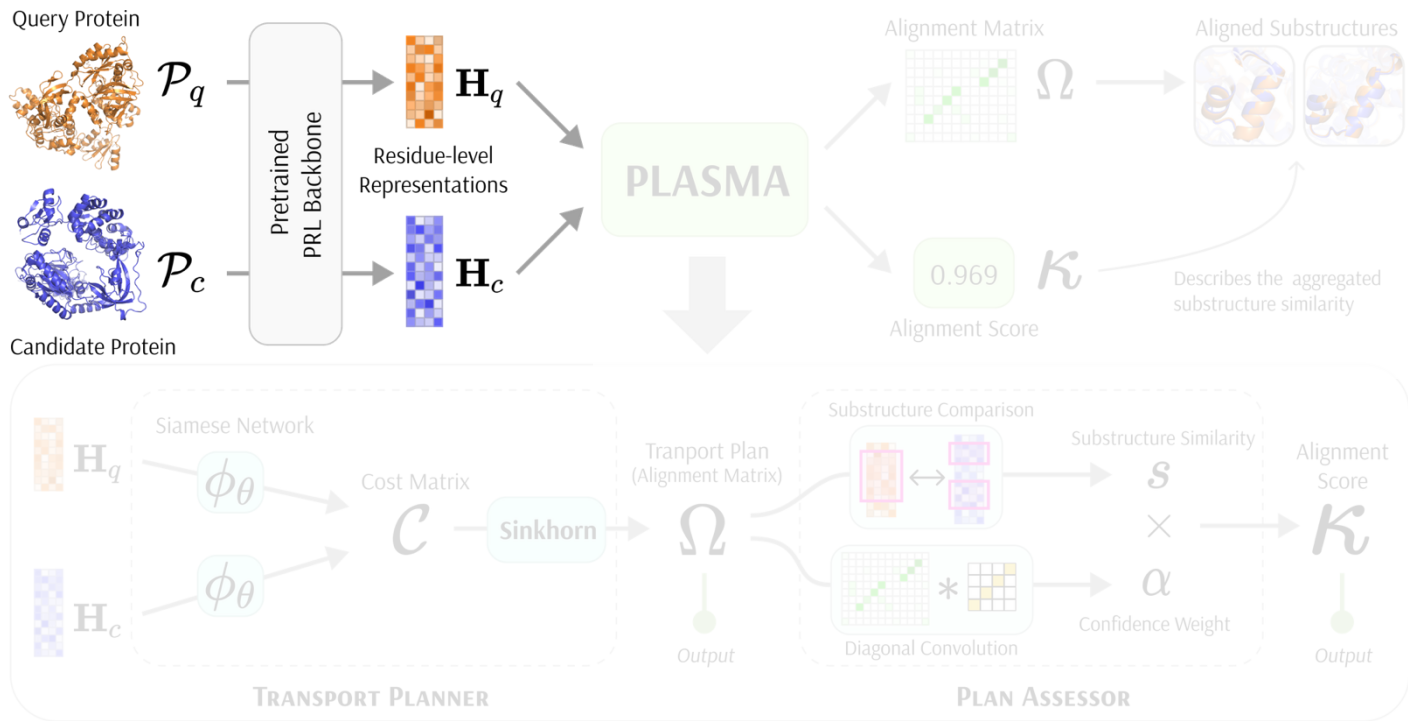
**~10ms per pair, 50× faster than TM-Align**



Poor local interpretability

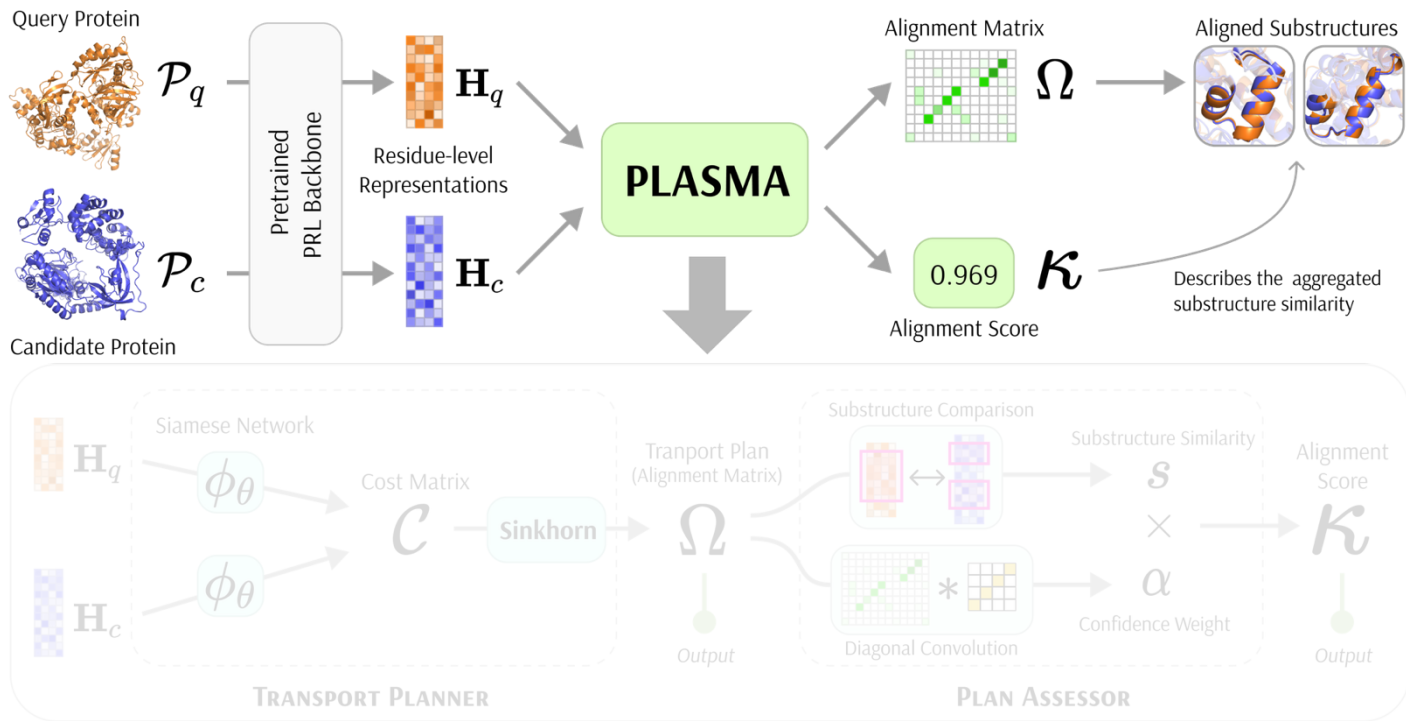
**Clear alignment matrices with bounded scores  $\kappa \in [0, 1]$**

# Step 1: Protein Embedding



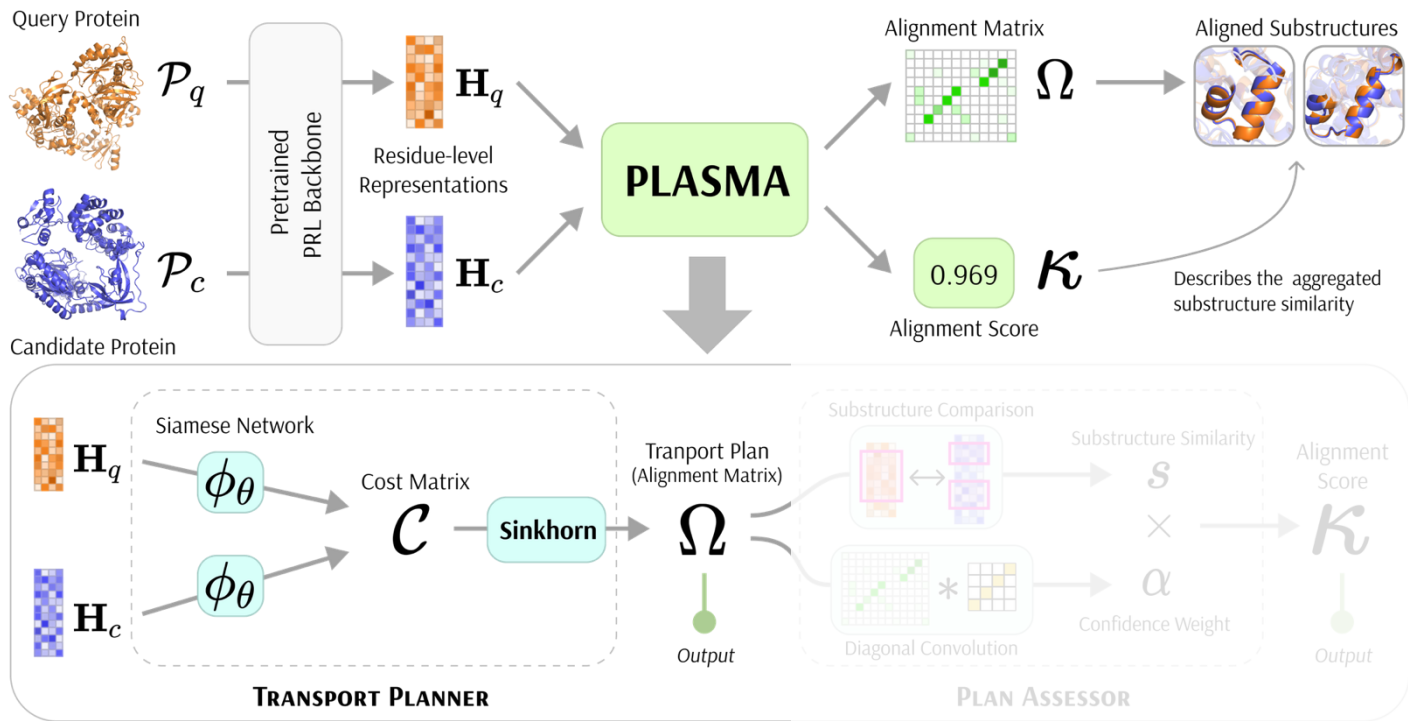
Given two proteins, we use a pretrained protein representation model to obtain residue-level embeddings.

## Step 2: PLASMA – Overview



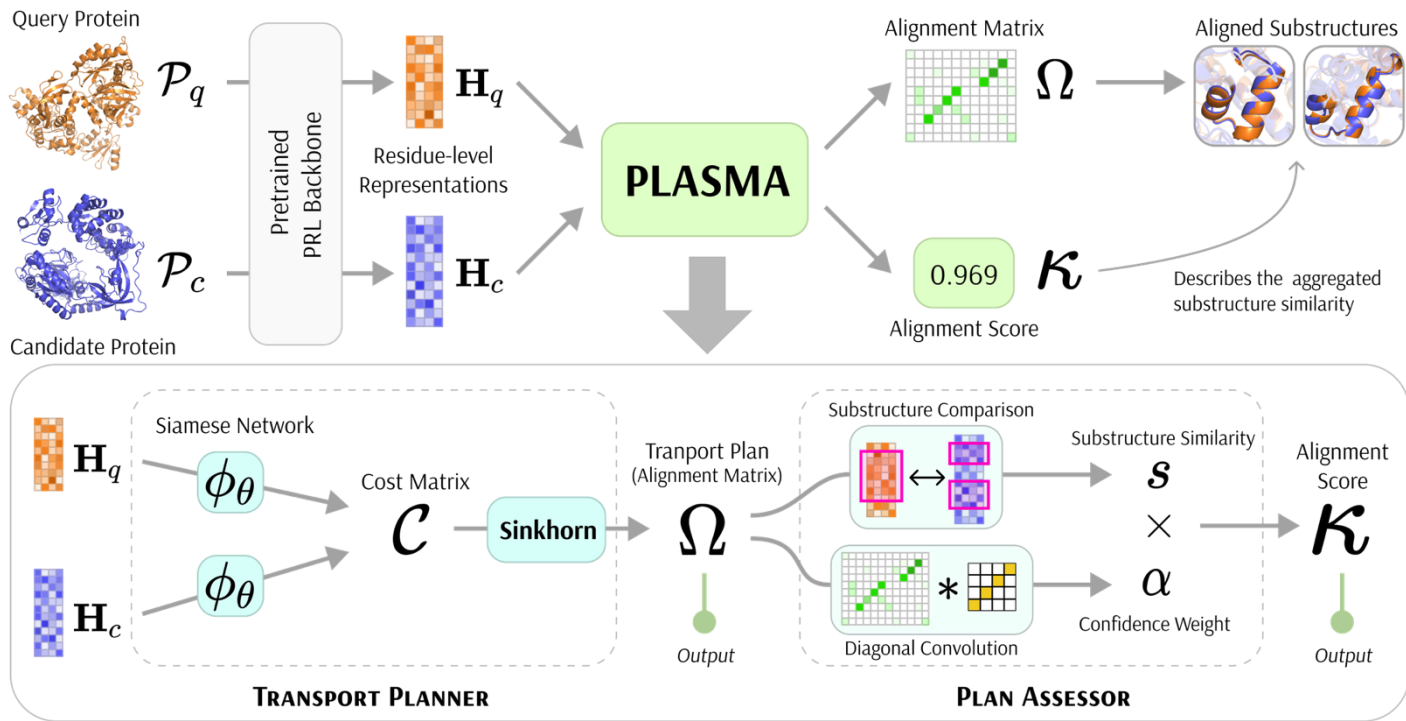
PLASMA takes the embeddings and returns an alignment matrix  $\Omega$  and a bounded similarity score  $\kappa \in [0, 1]$ .

# Step 3: Transport Planner



A siamese network learns the cost matrix  $\mathcal{C}$ , then Sinkhorn iterations produce the **alignment matrix  $\Omega$**  — this is the transport plan.

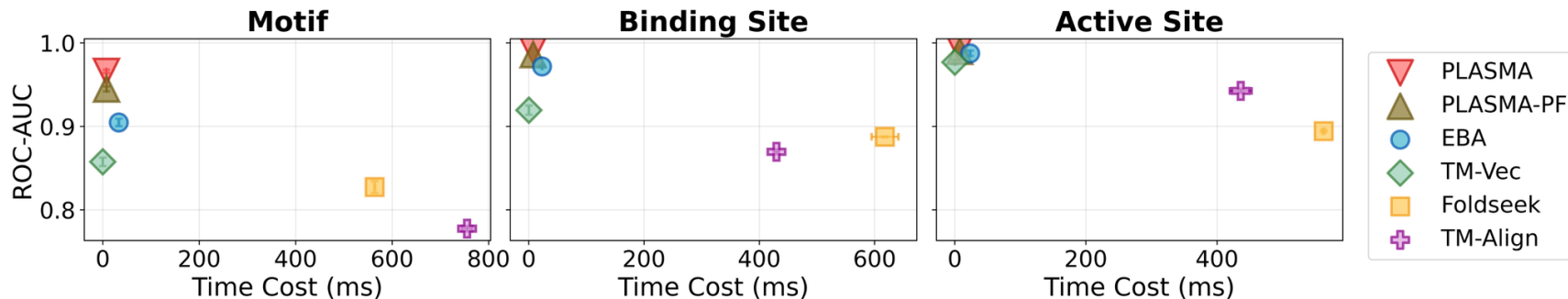
# Step 4: Plan Assessor



The Plan Assessor scores the alignment: substructure similarity  $\mathbf{s} \times$  confidence weight  $\alpha \rightarrow$  final bounded score  $\mathbf{k} \in [0, 1]$

# Evaluation

Faster and more accurate

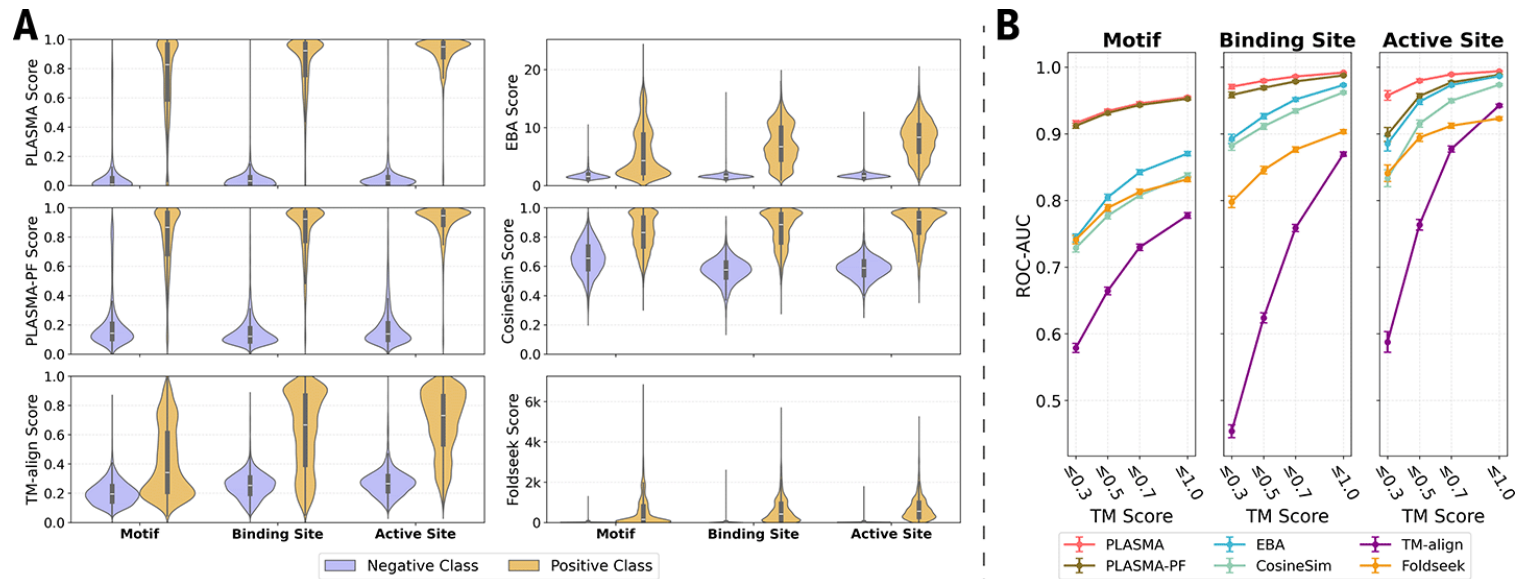


**PLASMA achieves the best ROC-AUC while running at ~10ms per pair**

50× faster than TM-Align/Foldseek, 3× faster than EBA

# Evaluation

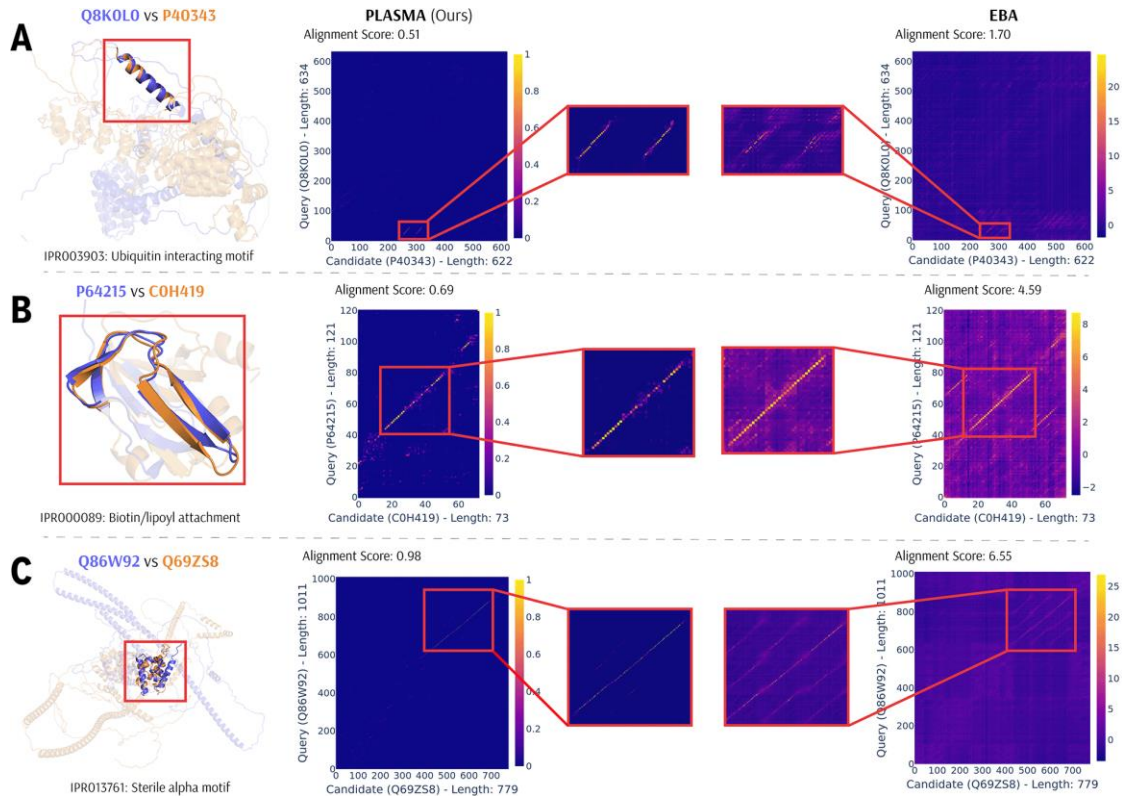
## Robustness across global similarity levels



PLASMA maintains ROC-AUC > 0.9 even when global similarity is very low (TM-score < 0.3)

# Evaluation

## Biological case studies



**PLASMA produces much cleaner alignments than EBA, focusing only on functionally important regions.**

# Thank You!

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**OpenReview**



[openreview.net/forum?id=FileqNzZzn](https://openreview.net/forum?id=FileqNzZzn)

**GitHub**



[github.com/ZW471/PLASMA-Protein-Local-Alignment](https://github.com/ZW471/PLASMA-Protein-Local-Alignment)