

PoinnCARE: Hyperbolic Multi-modal Learning for Enzyme Classification

Kun Xie¹, Peng Zhou², Xingyi Zhang³, Wei Liu¹,
Peilin Zhao⁴, Sibow Wang⁵, Biaobin Jiang¹

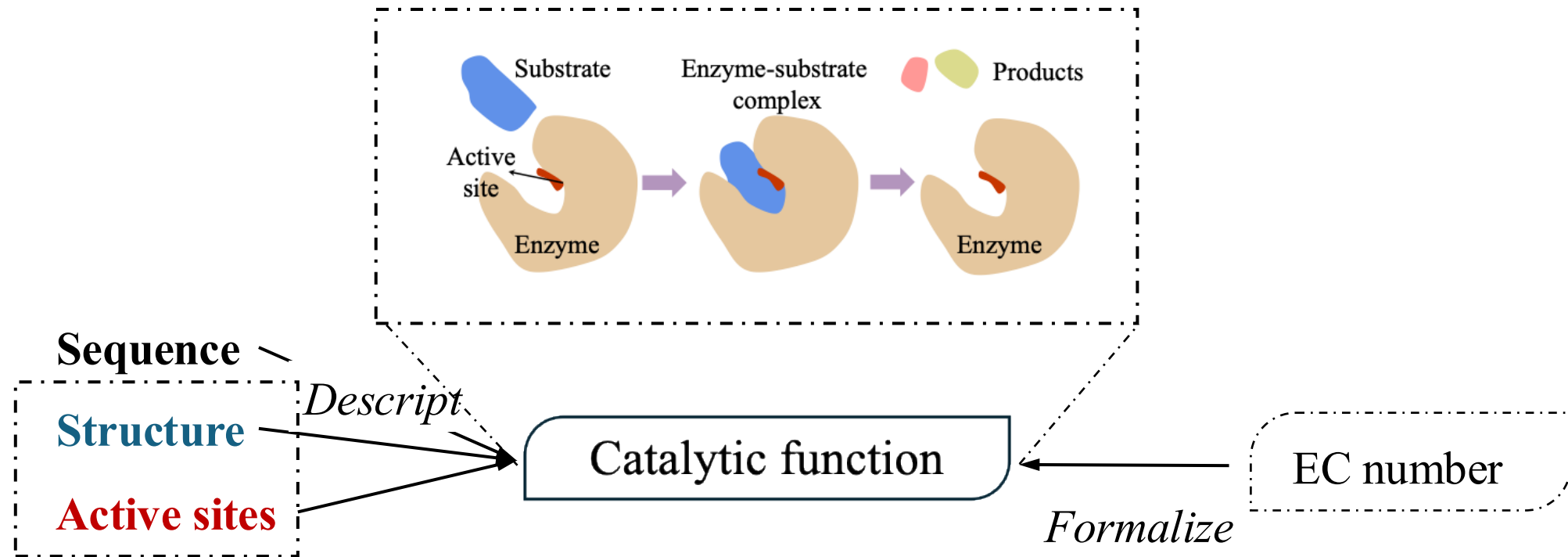
¹Tencent, ²Hunan University, ³MBZUAI,

⁴Shanghai Jiao Tong University, ⁵The Chinese University of Hong Kong

Presenter: XIE Kun

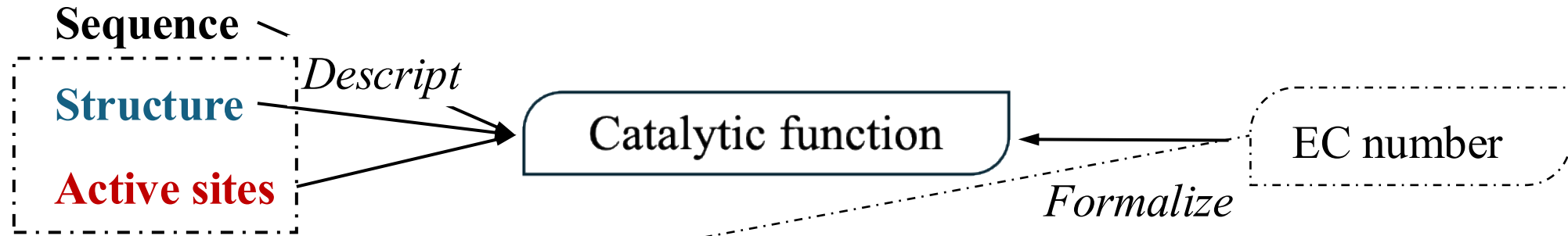
Enzyme & EC number

- Enzyme:
 - a protein catalyst
 - accelerates chemical reactions

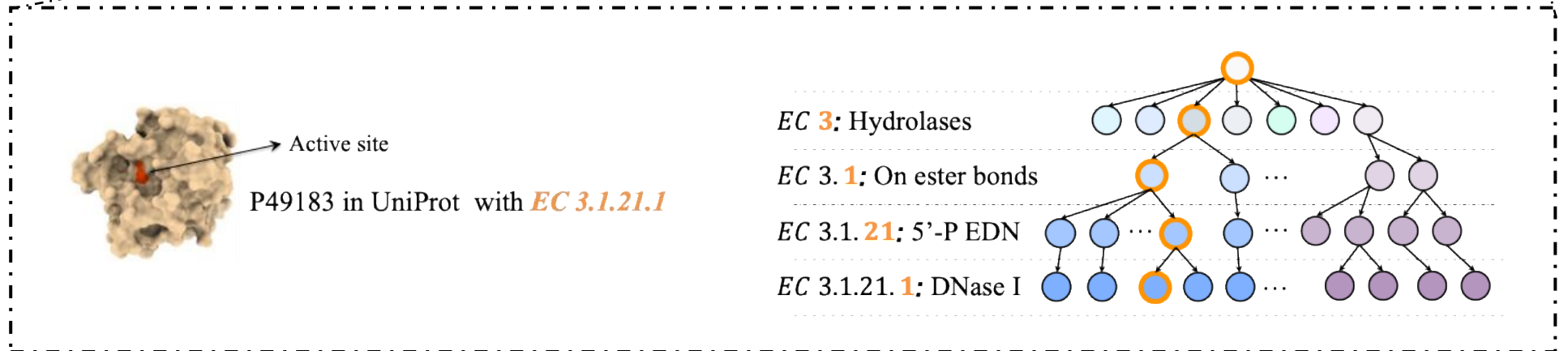


Important but *ignored* by existing methods

Enzyme & EC number

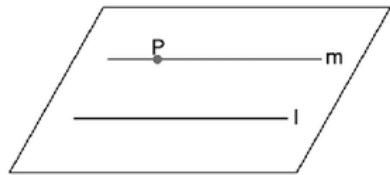


Important but *ignored* by existing methods

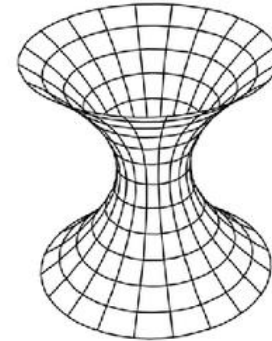


EC Number Prediction

- Input: Enzyme
- Output: EC number (\approx catalytic function)
- Challenge: ***Tree-like*** relationships among enzymes
 - Layer sizes grow ***exponentially***



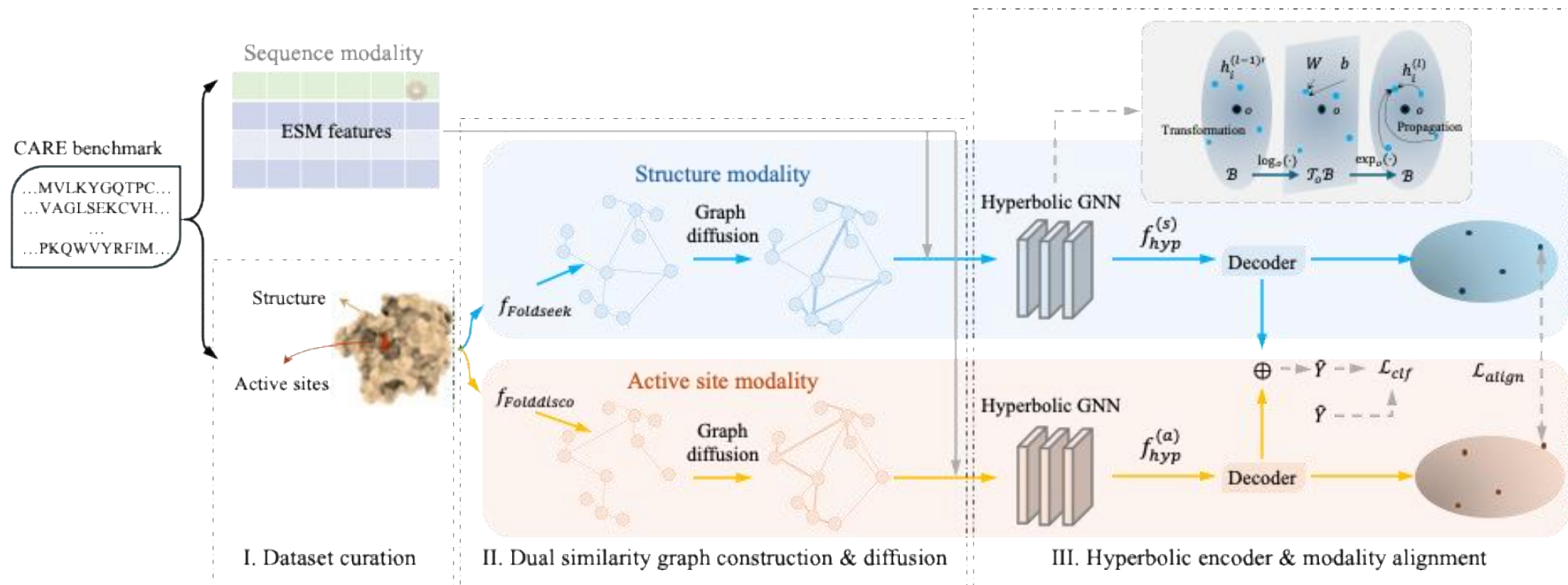
Euclidean space
In existing methods



Hyperbolic space
Efficient and accurate

Method: PoinnCARE

- 1. Learning in hyperbolic space
 - GNN in Euclidean space \rightarrow GNN in hyperbolic space

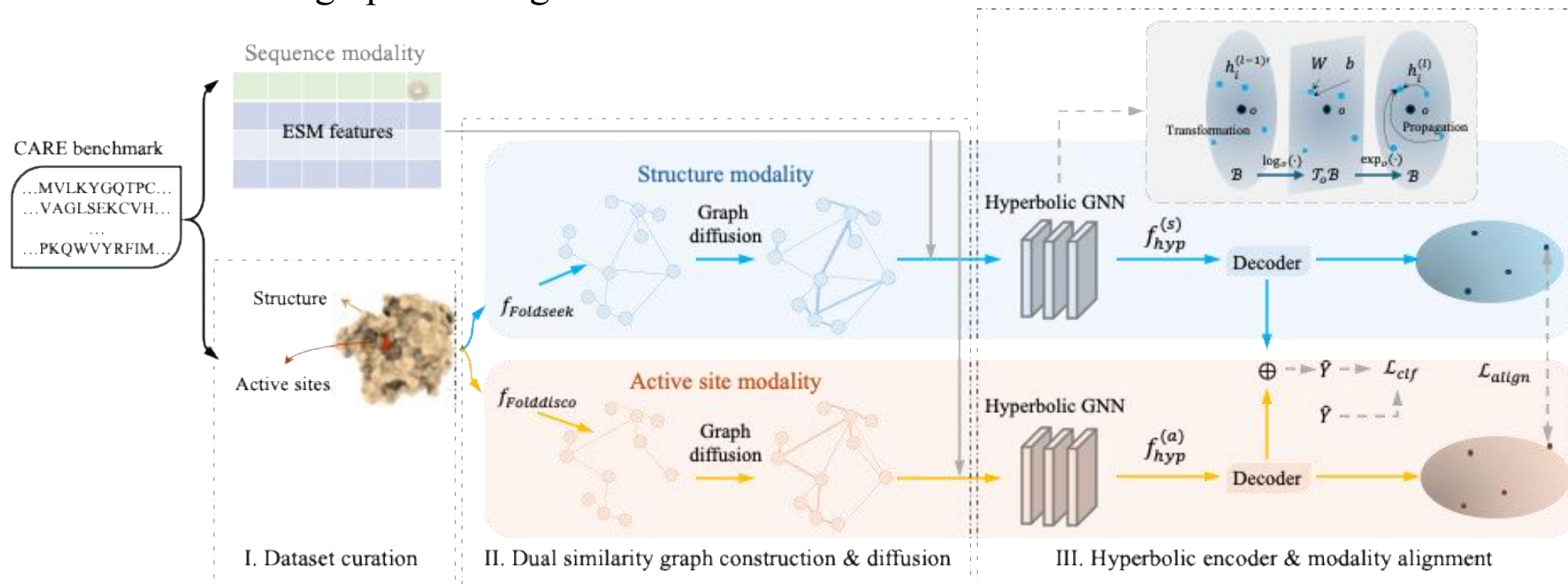


Method: Poinn CARE

- 2. Learning from multiple modalities
 - Dual similarity graphs for each modality, s.t.,
 - Graph diffusion to alleviate data sparsity
 - Multi-view graph learning to fuse different information



Sparse active site annotations
 → notable gap between modalities



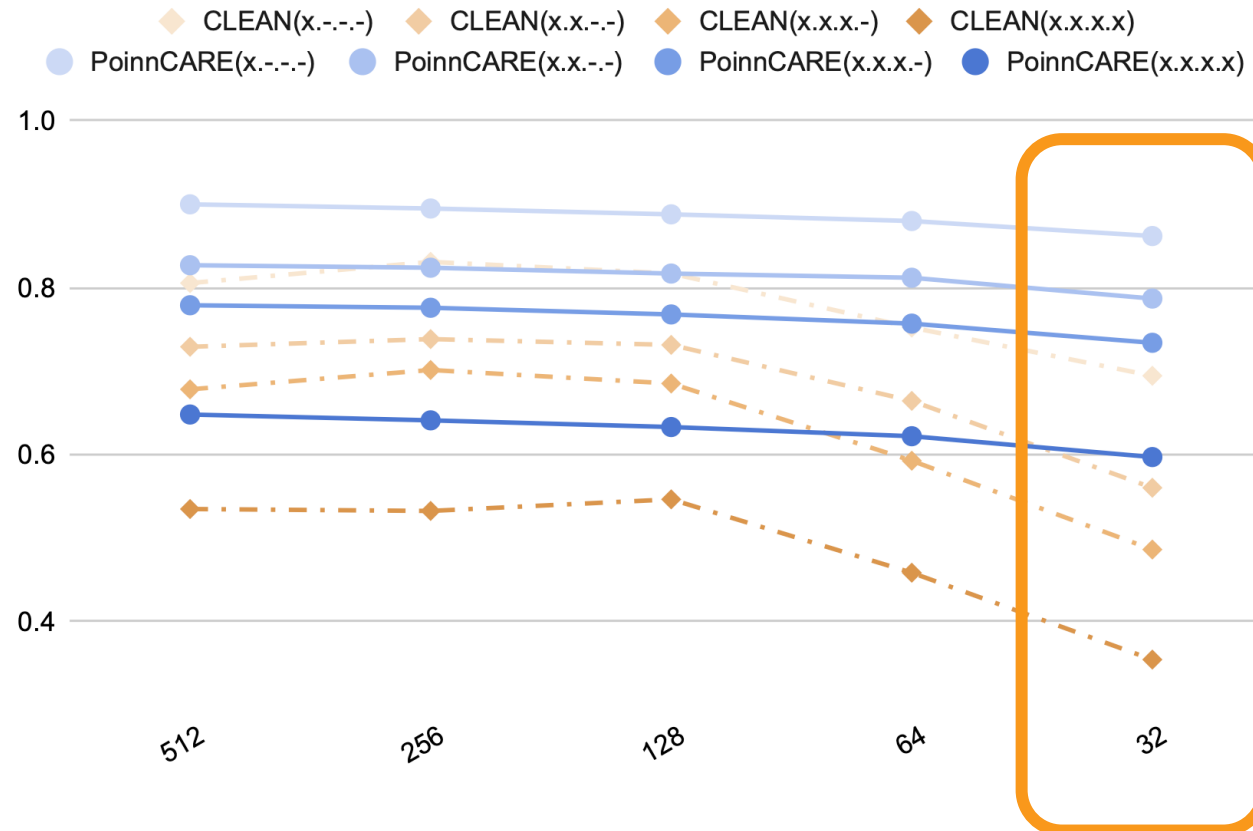
Experimental Results

- Systematic benchmark: CARE
- A broad range of SOTA baselines
- Comprehensive evaluation metrics

	<30% Identity				30-50% Identity				Avg. rank
	Level 1 (x.-.-)	Level 2 (x.x.-)	Level 3 (x.x.x.)	Level 4 (x.x.x.x)	Level 1 (x.-.-)	Level 2 (x.x.-)	Level 3 (x.x.x.)	Level 4 (x.x.x.x)	
Random*	0.194	0.032	0.012	0.000	0.225	0.036	0.007	0.000	13.38
BLASTp	0.697	0.590	0.569	0.475	0.923	0.879	0.850	0.773	7.00
Foldseek	<u>0.815</u>	0.722	0.667	<u>0.544</u>	0.932	0.880	0.841	0.755	4.13
Folddisco	<u>0.756</u>	0.600	0.511	<u>0.378</u>	0.798	0.755	0.723	0.564	9.88
CLEAN	0.806	<u>0.729</u>	<u>0.678</u>	0.535	0.946	<u>0.905</u>	<u>0.870</u>	<u>0.798</u>	2.50
CLEAN-Concat	0.810	<u>0.704</u>	<u>0.646</u>	0.507	<u>0.946</u>	<u>0.893</u>	<u>0.859</u>	<u>0.777</u>	3.63
ESM-2	0.783	0.695	0.643	0.518	0.944	0.895	0.856	0.781	4.13
ESM-c	0.691	0.574	0.527	0.436	0.911	0.848	0.808	0.745	9.25
ProtT5	0.755	0.649	0.604	0.492	0.929	0.873	0.833	0.765	6.38
ProtBert	0.672	0.546	0.502	0.410	0.874	0.804	0.767	0.709	10.38
S-PLM	0.751	0.637	0.582	0.470	0.921	0.861	0.823	0.752	7.75
ChatGPT*	0.278	0.016	0.000	0.000	0.336	0.030	0.014	0.000	13.38
Pika*	0.616	0.461	0.377	0.206	0.738	0.600	0.502	0.377	12.00
PoinnCARE	0.900	0.827	0.779	0.648	0.961	0.926	0.887	0.822	1.00

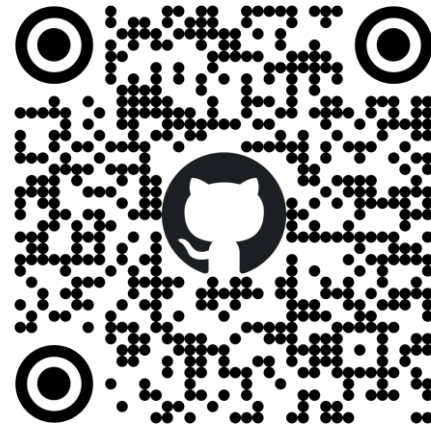
Experimental Results

- Efficient & Accurate
- High accuracy, even under 32 dimension



Conclusion

- Enhanced benchmark
 - += structural info
 - += active site info
- Similarity graphs
 - → Graph diffusion
- Hyperbolic
 - Multi-modality encoding
 - Multi-modality alignment
- Extensive experiments



Learn more  **ICLR**