TrustAffinity: accurate, reliable and scalable out-of-distribution protein-ligand binding affinity prediction using trustworthy deep learning

Amitesh Badkul¹, Li Xie¹, Shuo Zhang^{1,3}, Lei Xie^{1, 2, 3} ¹Hunter College, City University of New York ²The Graduate Center, City University of New York ³Weill Cornell Medicine, Cornell University



Motivation

Despite, the integration of AI for accelerating drug discovery, the existing algorithms fail:

- To **generalize** for understudied proteins or compounds from unlabeled protein families or chemical scaffolds.
- To quantify uncertainty associated with their predictions.
- To achieve scalability to billions of compounds.
 Overcoming these challenges is key to successfully integrating machine learning with drug discovery with higher efficacy and safety.

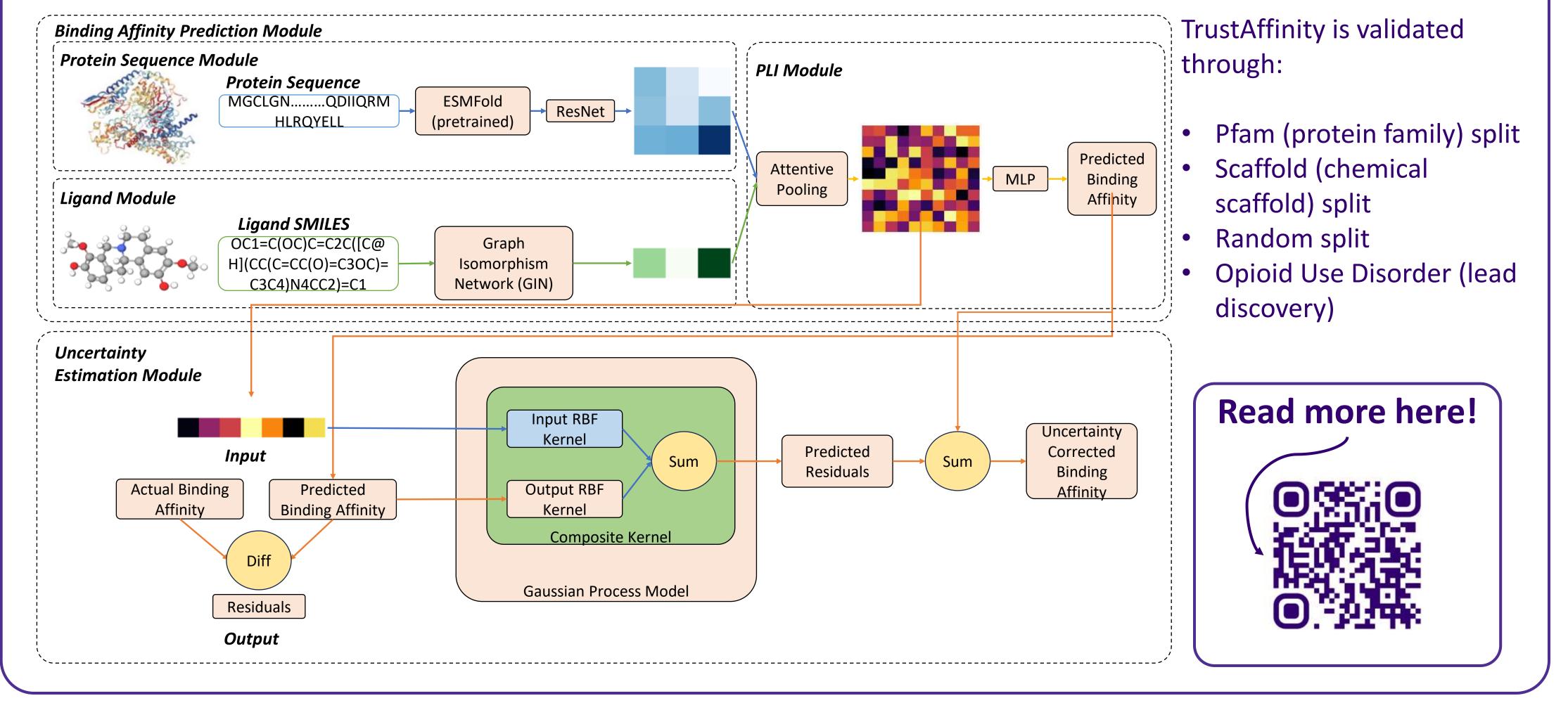
Contributions

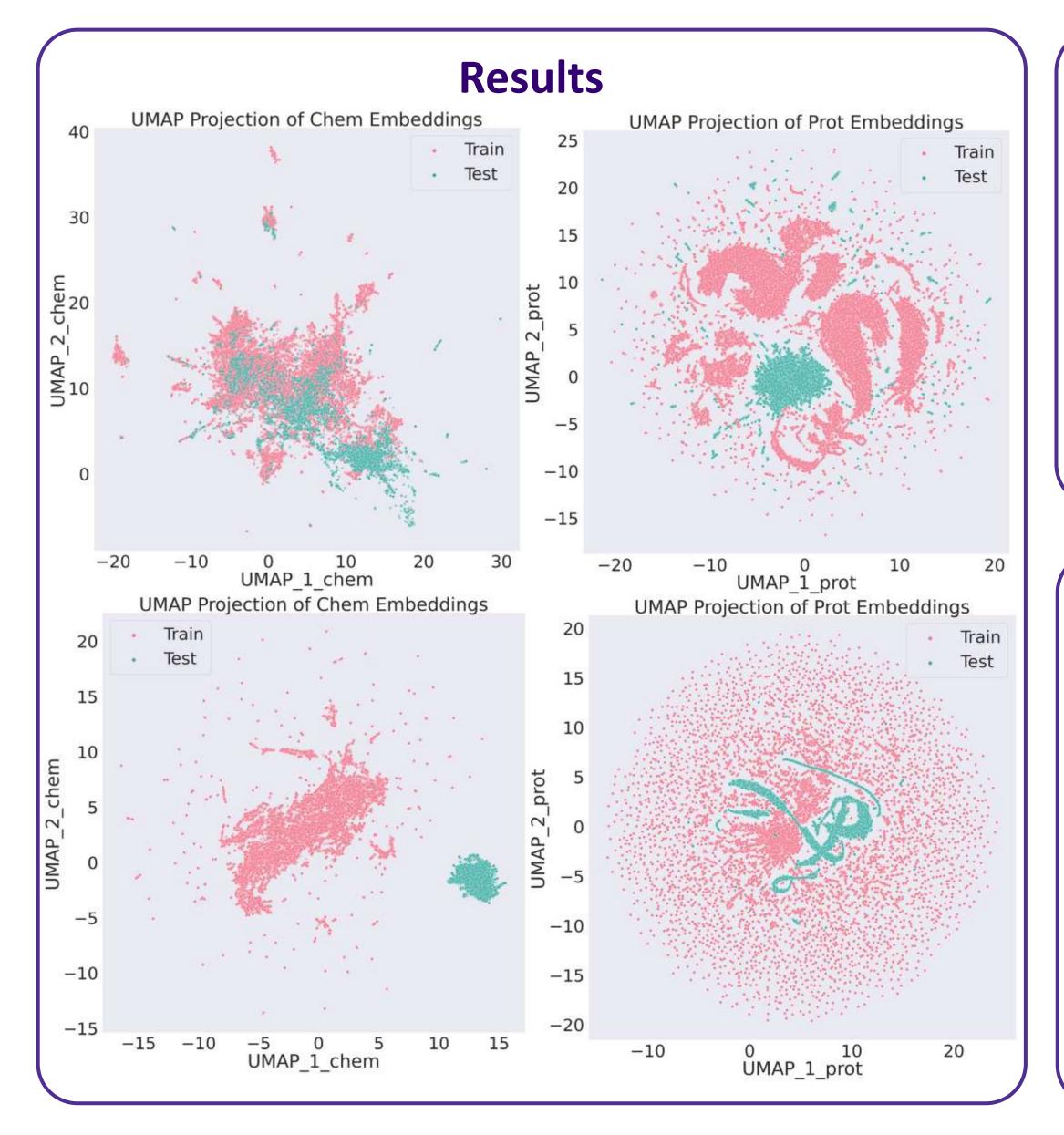
We address these limitations by introducing **TrustAffinity**:

- A **sequence based** deep learning framework with efficient uncertainty quantification based on residue-estimation.
- Rigorously validated TrustAffinity across various OOD scenarios, surpassing SOTA methods.
- We demonstrate the practicality of our framework through **Opioid Use Disorder lead discovery**.

These are vital for the successful incorporation integrating machine learning with drug discovery.

Method





Opioid Use Disorder (OUD) Results						
Method	RMSE	MAE	r	ρ		

AutoDock Vina	1.179	1.031	0.308	0.334
BACPI	2.523	2.181	0.103	0.122
TrustAffinity (Y_{true})	0.384	0.312	0.856	0.820
TrustAffinity (Y_{pred})	0.846	0.65	0.612	0.667

Conclusion

TrustAffinity successfully:

- Displays ability to differentiate in Pfam and Scaffold OOD scenarios.
- Achieves high confidence performance when compared to machine learning and docking based methods.
- Exhibits highest performance for OUD lead discovery case.
- Runs at **three orders** of magnitude faster than traditional protein-ligand docking.