

# Binding Affinity Prediction of Protein-Ligand complexes using Machine Learning

## MSc Project

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What are a proteins and ligands?

- **Proteins:** Complex molecules that are work-horses (machines) of a living organism.
- **Ligands:** Molecules that bind to particular proteins, called receptor proteins.
- Proteins and ligands bind together to form protein-ligand complexes.

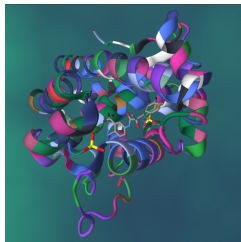


Figure: Haemoglobin transporter protein [6].



## Protein-Ligand complexes

- Any potential binding location in the 3D structure of a protein is called a pocket.
- The pockets of proteins only bind to ligands of complementary shape.
- Drugs are just ligand molecules that bind to protein to cause a therapeutic effect.

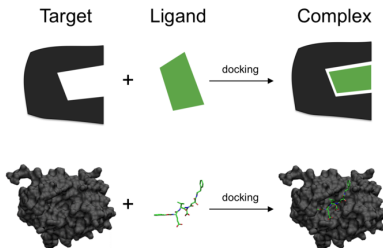


Figure: Lock and Key hypothesis in molecular docking [7].

## Understanding Protein-Binding Affinity.

- Binding affinity between a protein and a ligand is quantified by the  $K_d$ ,  $K_i$  and  $IC_{50}$ . Here  $K_d$  refers to the dissociation constant,  $K_i$  to inhibition constant, and  $IC_{50}$  to inhibitory concentration 50%.
- $K_d$  can be quantified by using protein concentration  $[P]$  and ligand concentration  $[L]$  at equilibrium  $[1]$ .

$$K_d = \frac{[P][L]}{[PL]}$$

- $K_i$  and  $IC_{50}$  are similarly defined.



# Problem Definition

- Determining if a potential drug (ligand) can bind to a target protein is very costly processes [2].
- The project aims to predict the ligand affinity based on previously recorded data ("In-Silico" method). This reduces the drug discovery costs.
- We use PDB databank, which holds PL affinity data collected over many decades.

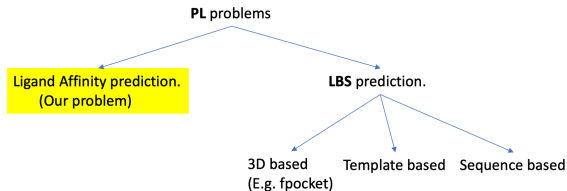


Figure: Protein-Ligand problem classification.



**PDB databank** (v2019) was used to extract input features.

- We use *fpocket/dpocket* ligand binding site prediction library to get the features of pockets pockets in proteins.
- *RDKit* library is used to extract features for each ligand.
- **Ligand Features:** Using *RDKit.Chem.Descriptors*, 402 features were extracted for each ligand. Hence the ligand features space was  $\mathbf{R}^{402}$ .
- **Protein Features:** For every pocket, 55 descriptors are obtained in total. Hence, the input space for protein features is  $\mathbf{R}^{55}$

The concatenated input feature space before input feature elimination  $\mathbf{R}^{457}$ .



# Feature Selection

We only had 16000 data points to train a feature space of  $\mathbf{R}^{457}$ . We reduced our features using the following feature selection strategies:

- **Output Correlation:** The input features that have the best *Pearson* and *Spearman* correlation were selected. [5].
- **Genetic Algorithms:** Genetic algorithms with the following score function was used to select the best features [4]:

$$\text{score} = \mathbf{R}^2_{\text{score}} * \text{Features Eliminated}$$

- **Manual Feature Selection:** A selected list of 121 ligand descriptors was used with all protein descriptors as input to the model.





# Testing strategy

We use the following methods to determine the quality of results and reproducing them:

- **Reproducibility:** To reproduce the results, we use report random seed (Execution ID) for every execution.
- **$R^2$  score** (*Coefficient of determination*) [3]: .  $R^2 \in (-\infty, 1.0]$  where 1.0 is the best score.
- **Visualization:** Our model's approximated function  $f : \mathbf{R}^n \mapsto \mathbf{R}$  where  $n \in \mathbf{I}^+$  is visualized as a 2D scatter plot.

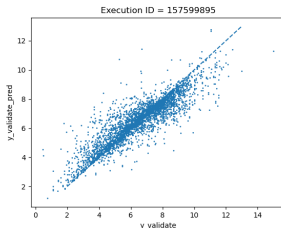









Figure: (Sample) Visualizing accuracy.  $R^2 \approx 0.805$ .



## Q & A



-  Du, Li, Xia, Ai, Liang, Sang, Ji and Liu; Insights into Protein–Ligand Interactions: Mechanisms, Models, and Methods (2016)
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-  scikit-learn; R2 score, the coefficient of determination,
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-  Protein–Ligand complex; Wikipedia.
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