



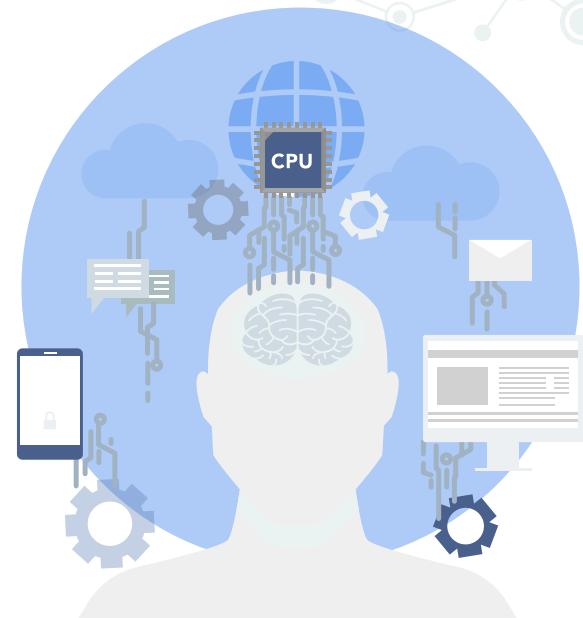
# GRADUATION THESIS

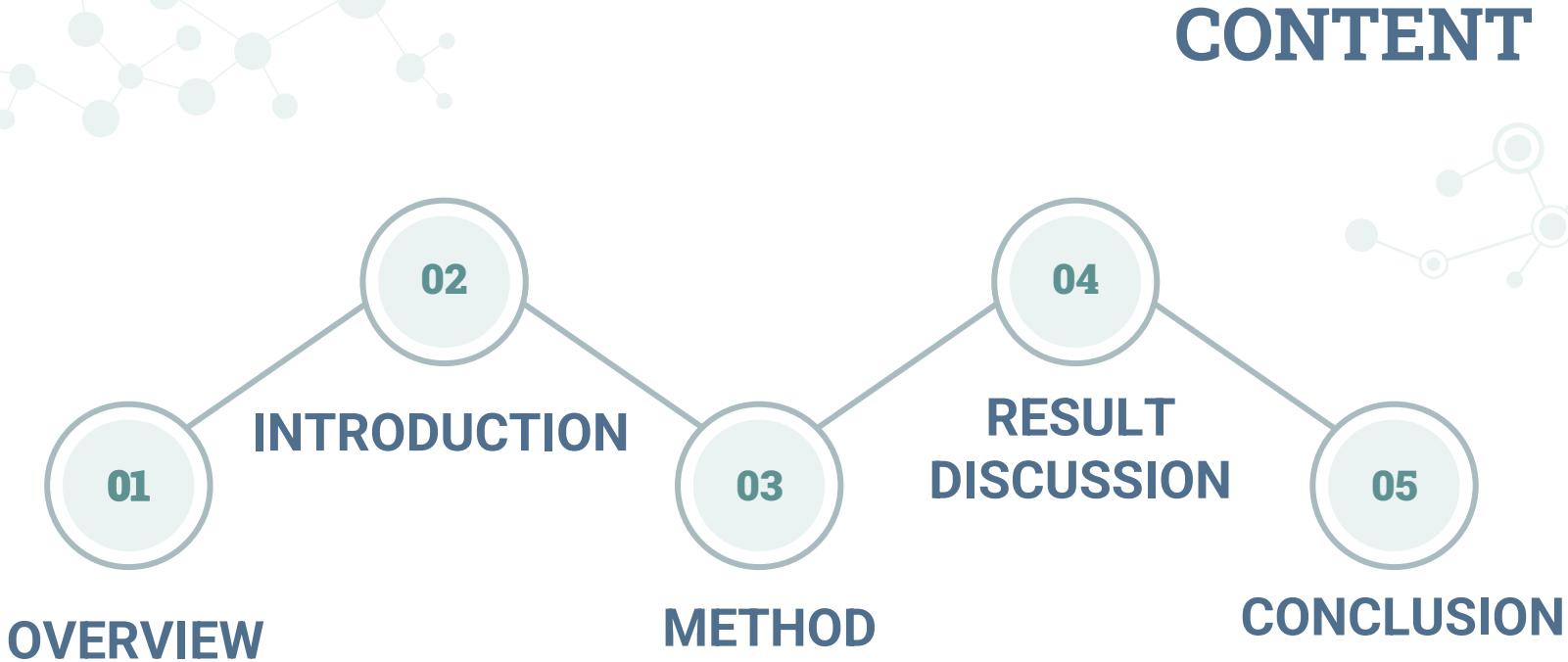
Department of Organic Chemistry

## *IN SILICO* MODELING FOR PREDICTION OF POTENTIAL HIV-1 INTEGRASE INHIBITORS

Presenter: Phan Tieu Long

Supervisor: Assoc.Prof.Truong Ngoc Tuyen

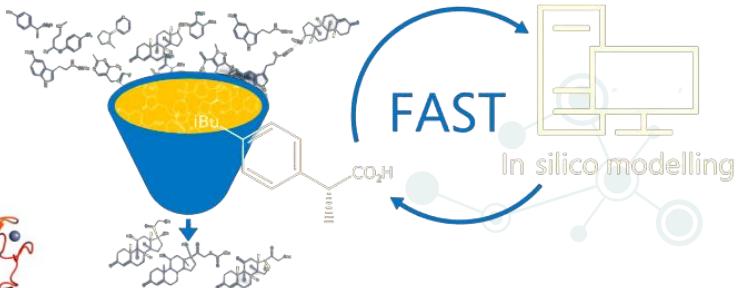
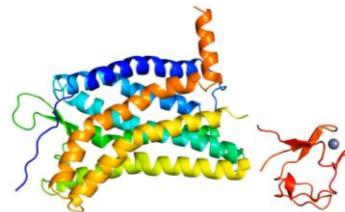




# OVERVIEW



# OVERVIEW





## OVERVIEW

MOLECULAR DOCKING

QSAR

PHARMACOPHORE

VIRTUAL  
SCREENING





## OVERVIEW

01

Pharmacophore

02

QSAR  
classification

03

QSAR  
regression

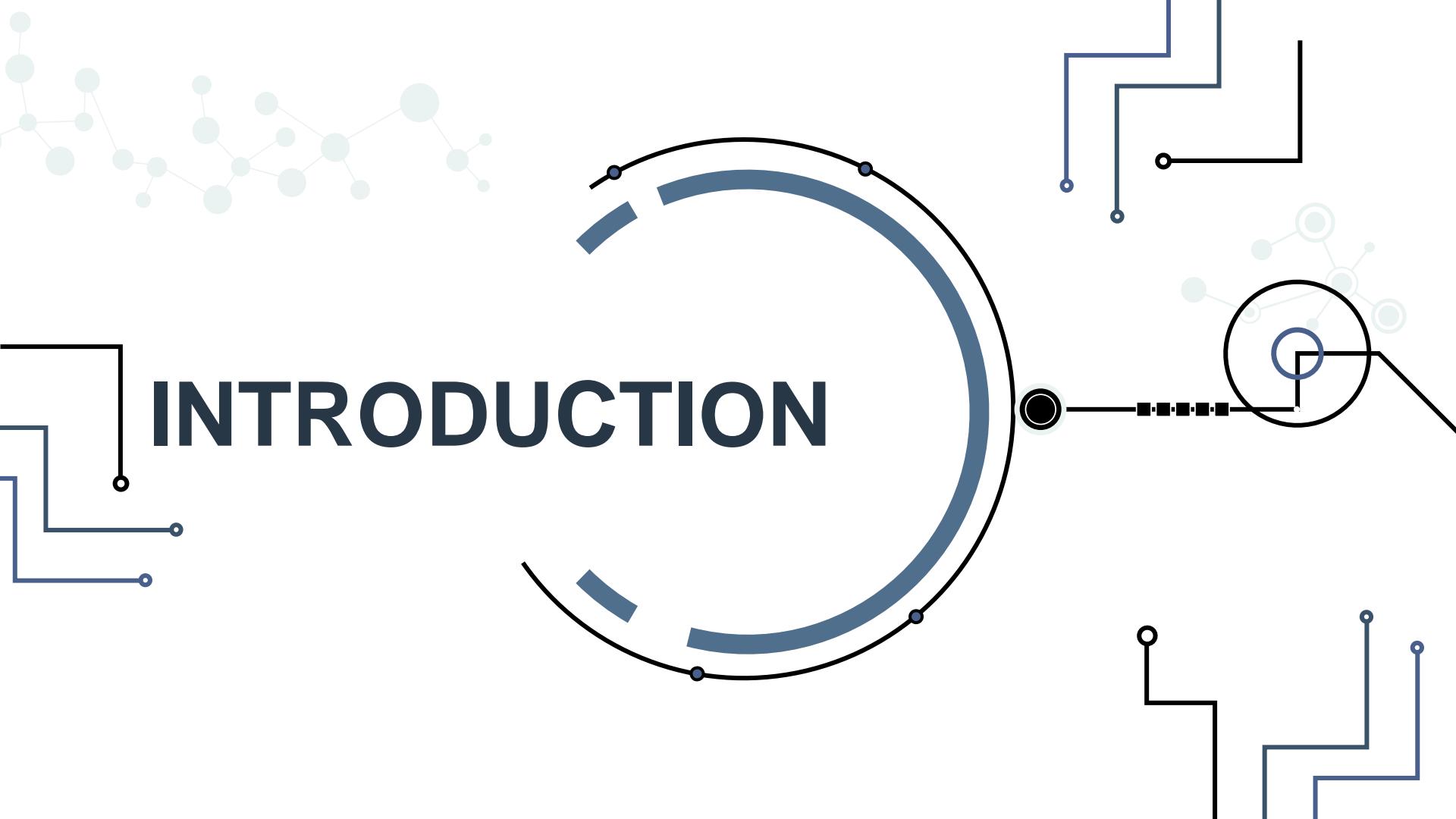
04

Docking

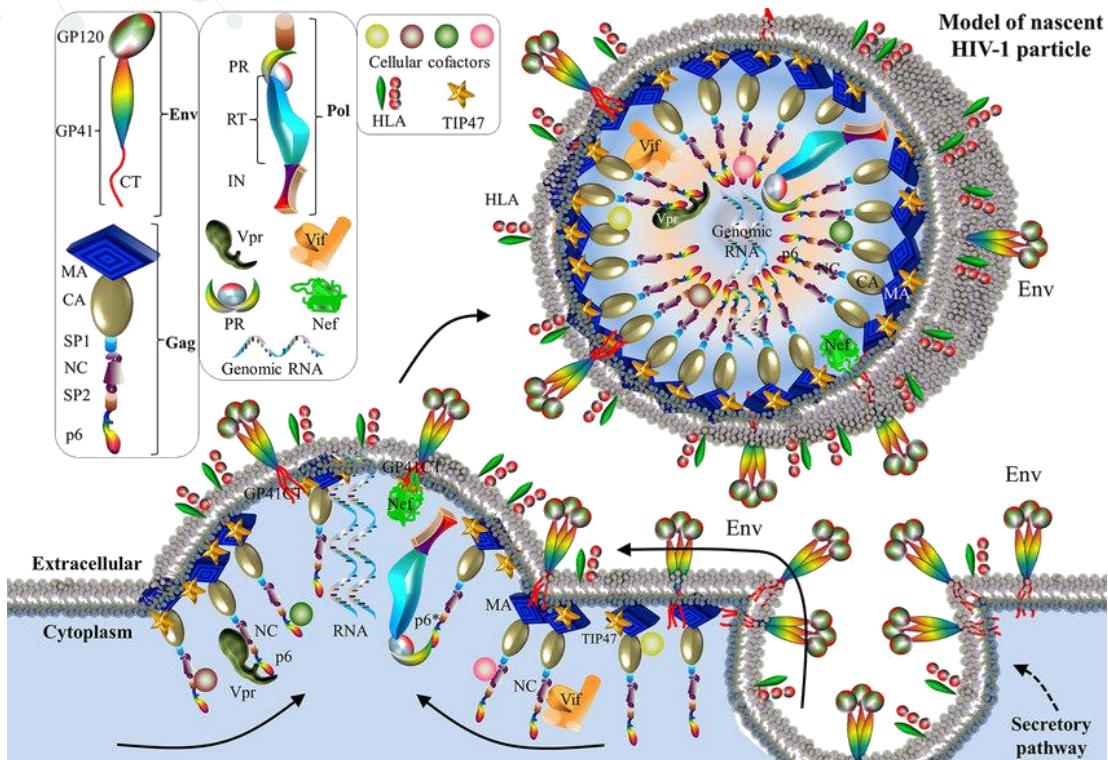
05

Screening

# INTRODUCTION

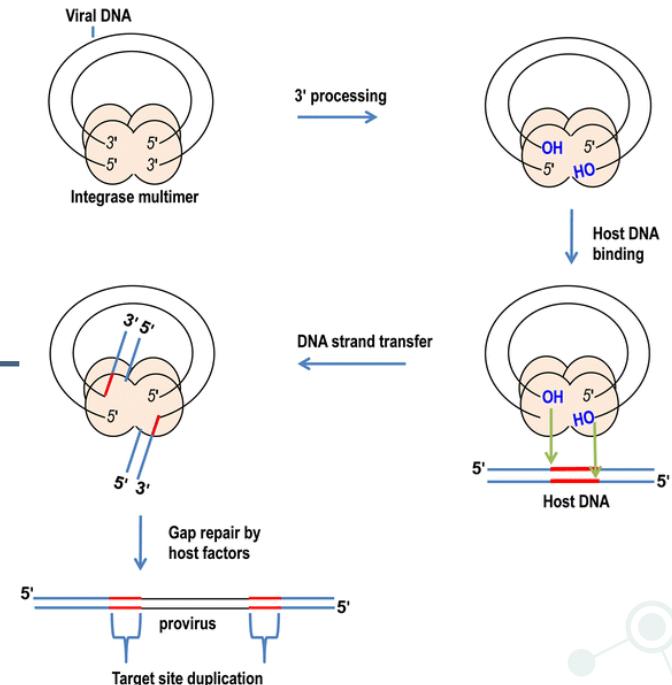
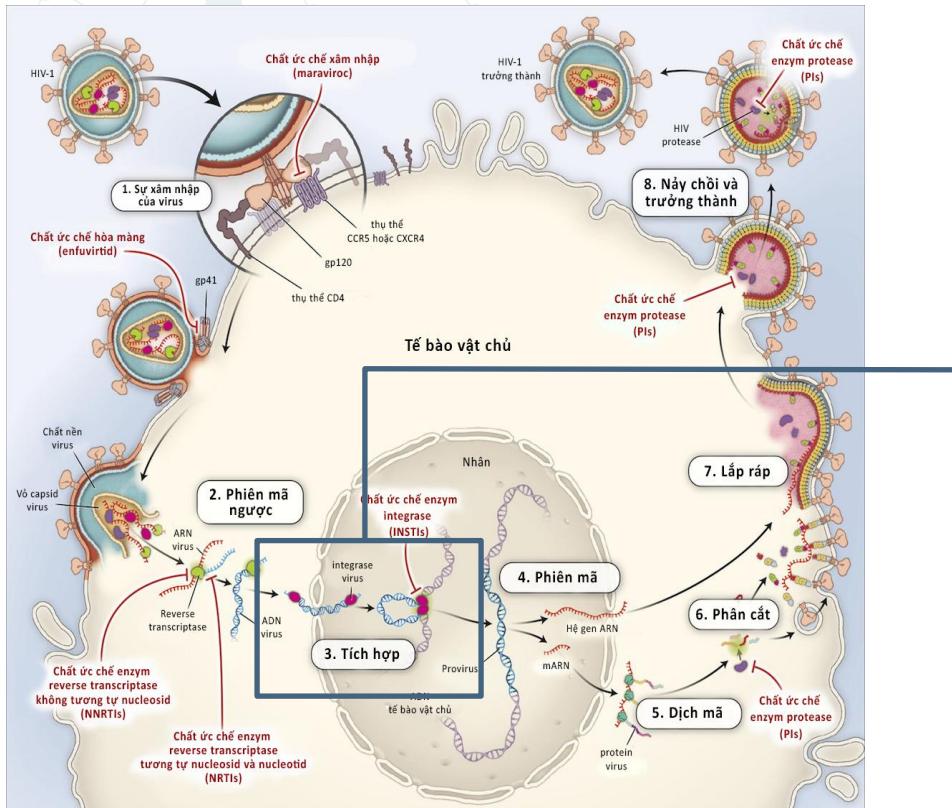


# INTRODUCTION



Li G, De Clercq E. HIV Genome-Wide Protein Associations: a Review of 30 Years of Research.  
Microbiology and molecular biology reviews : MMBR. 2016;80(3):679-731. doi:10.1128/mmbr.00065-15

# OVERVIEW

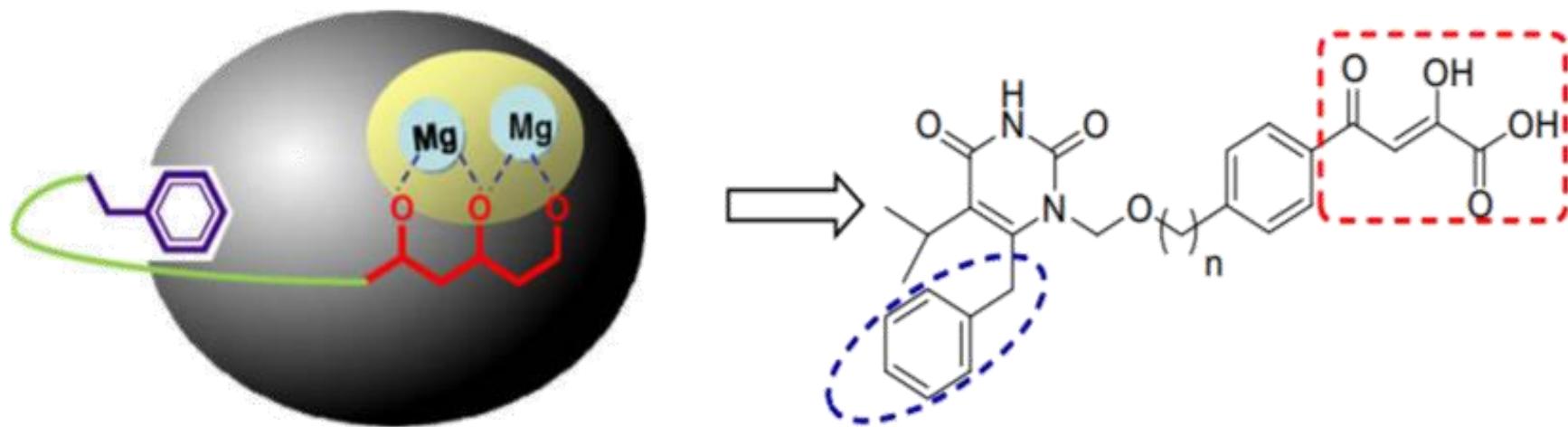


## INTEGRATION MECHANISM

Gandhi M., Gandhi R. T. J. N. E. J. o. M. (2014), "Single-pill combination regimens for treatment of HIV-1 infection".

## OVERVIEW

## SAR OF INSTIS



Wang Z., Tang J., Salomon C. E. et al. (2010), "Pharmacophore and structure-activity relationships of integrase inhibition within a dual inhibitor scaffold of HIV reverse transcriptase and integrase", *Bioorganic & Medicinal Chemistry*. 18(12), pp. 4202-4211

# OVERVIEW

# FDA APPROVED DRUGS

**Cabotegravir  
2021**



**Bictegravir  
2018**



**Dolutegravir  
2013**



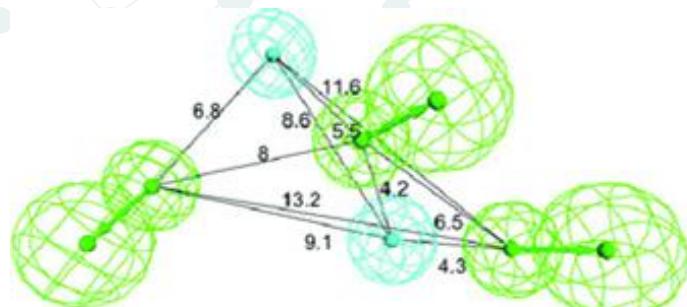
**Elvitegravir  
2012**



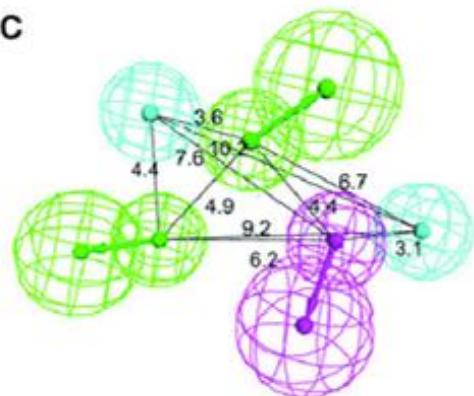
**Raltegravir  
2007**



## OVERVIEW

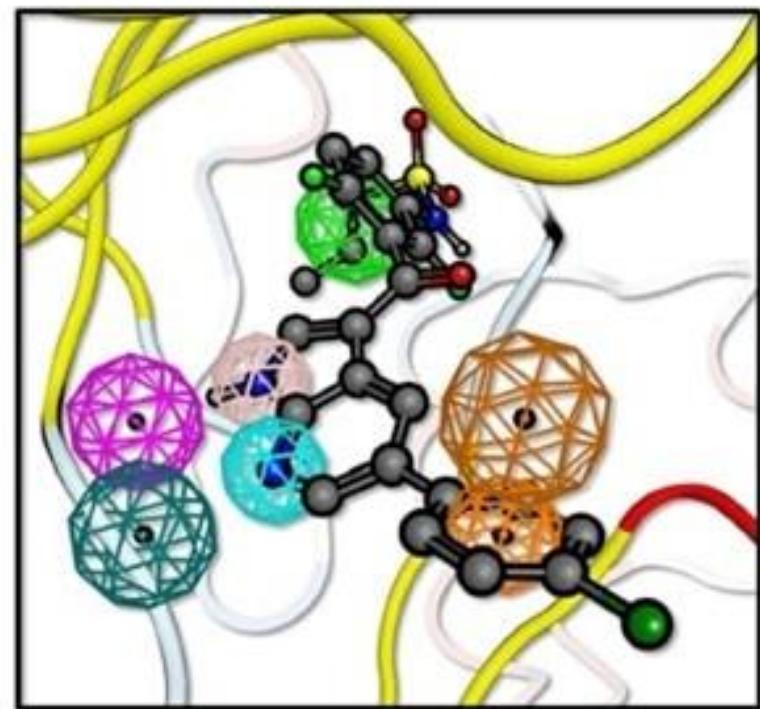


C



Ligand-based

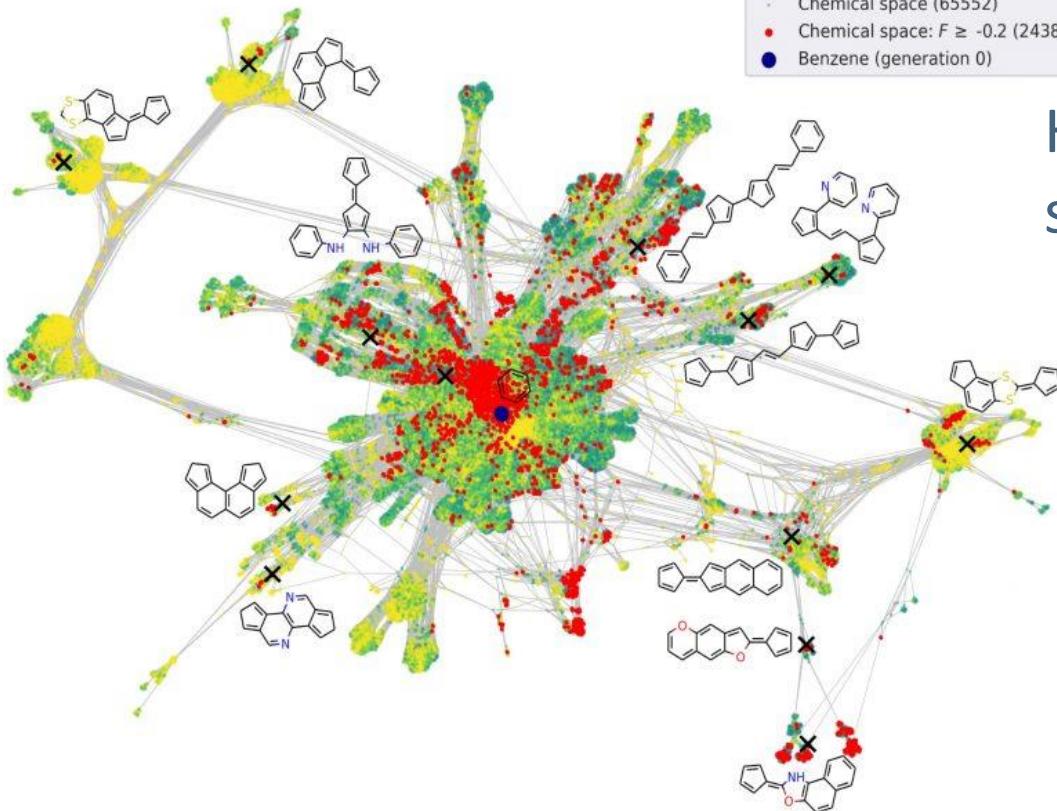
## PHARMACOPHORE



Protein-based

## OVERVIEW

## PHARMACOPHORE



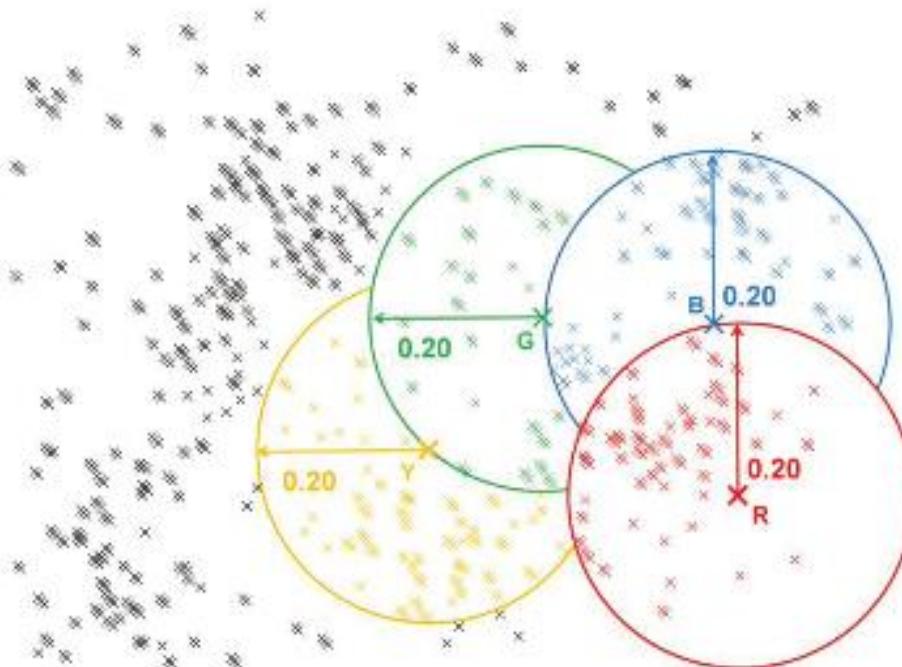
How can we select diverse subset for training model?



## OVERVIEW

## PHARMACOPHORE

### Butina algorithm

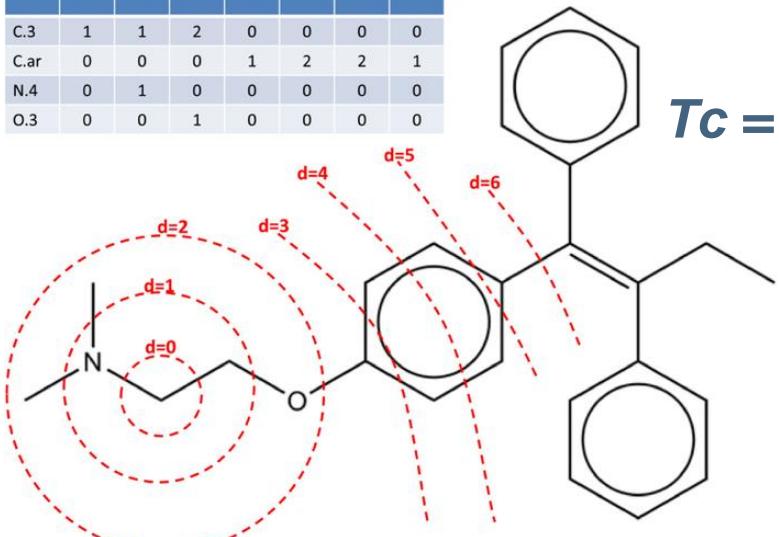


- Calculate similarity matrix
- Select centroid
- "Single Pass" technique

Downs GM, Barnard JM. Clustering methods and their uses in computational chemistry. *Reviews in computational chemistry*. 2002;18:1-40.

## OVERVIEW

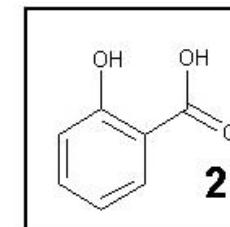
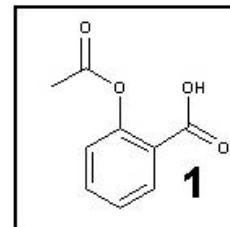
	d=0	d=1	d=2	d=3	d=4	d=5	d=6
C.3	1	1	2	0	0	0	0
C.ar	0	0	0	1	2	2	1
N.4	0	1	0	0	0	0	0
O.3	0	0	1	0	0	0	0



Rogers D, Hahn M. Extended-connectivity fingerprints. *Journal of chemical information and modeling*. 2010;50(5):742-54. doi:10.1021/ci100050t

## PHARMACOPHORE

### Molecular fingerprint



<b>1</b>	1	1	0	1	1	0	1	0
<b>2</b>	1	1	0	1	0	0	0	0

Bender A, Glen RC. Molecular similarity: a key technique in molecular informatics. *Organic & biomolecular chemistry*. 2004;2(22):3204-18. doi:10.1039/b409813g

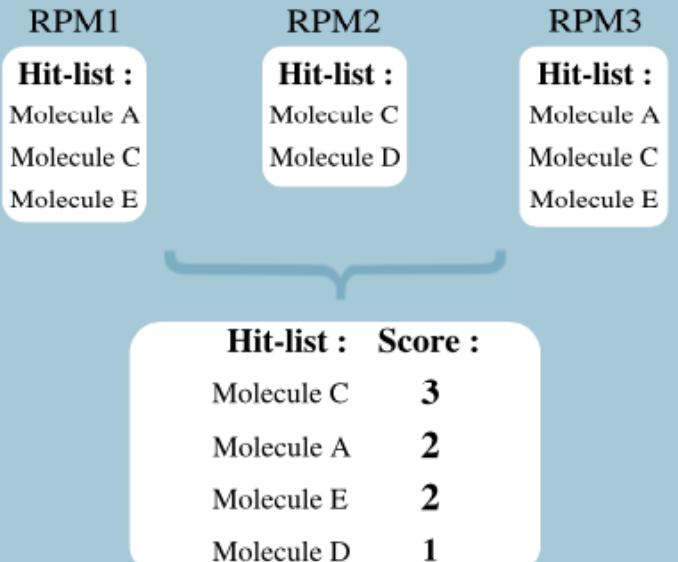


## OVERVIEW

## PHARMACOPHORE

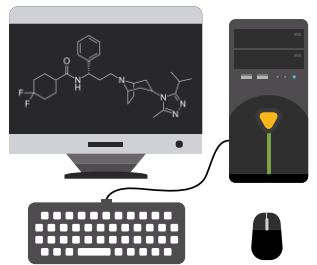
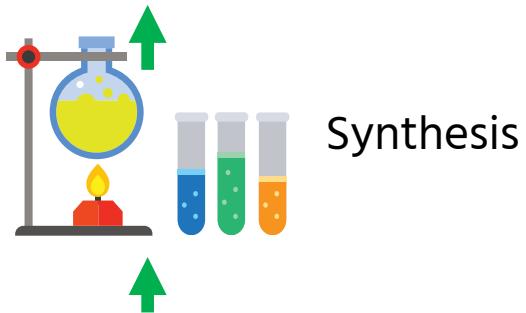
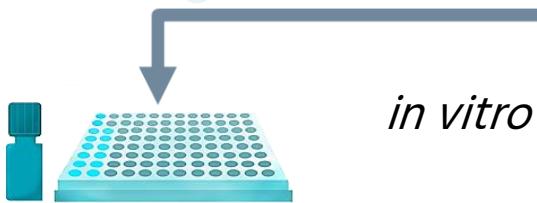
### Model optimization

#### Common Hits Approach

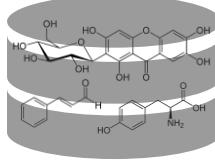


- Change query threshold
- Change overlap ratio
- Add exclusion volume
- Change ligand's shape
- **Common Hits Approach**

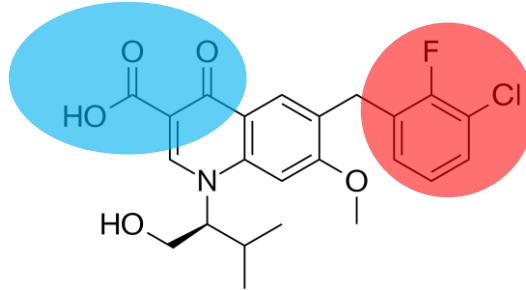
## OVERVIEW



## Database

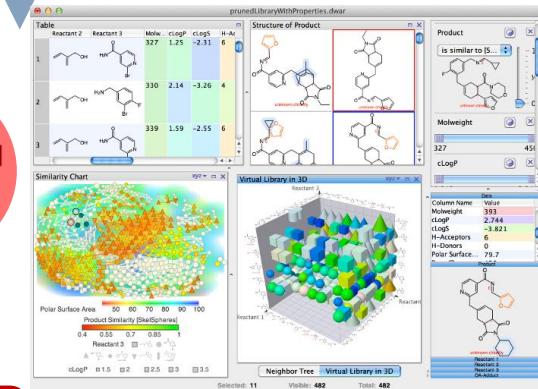


## QSAR



## QSAR MODEL

**Lead compound**



## OVERVIEW



### mordred-descriptor/ documentation

<http://mordred-descriptor.github.io/documentation/master>



1 Contributor    0 Issues    0 Stars    0 Forks



and Machine Learning

### phi-grib/PaDEL- descriptor-ws

PaDEL ws descriptors engine

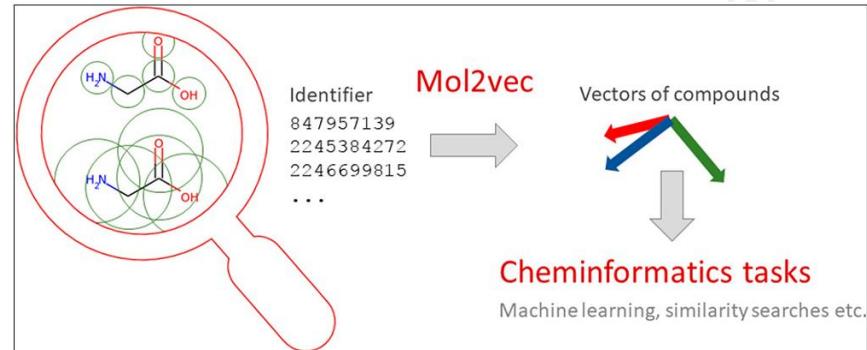


1 Contributor    1 Issue    11 Stars    6 Forks



## QSAR MODEL

### Molecular representation

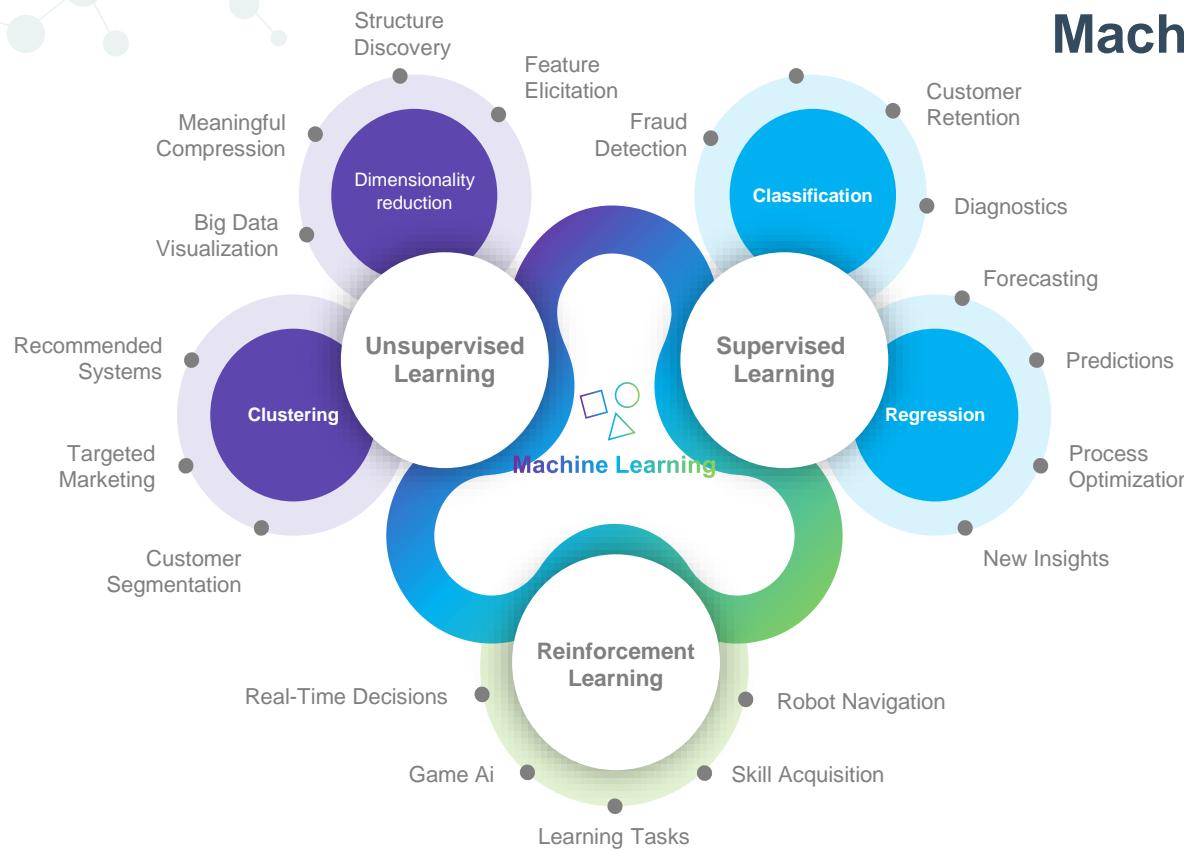


Open-Source Cheminformatics  
and Machine Learning

## OVERVIEW

## QSAR MODEL

# Machine Learning



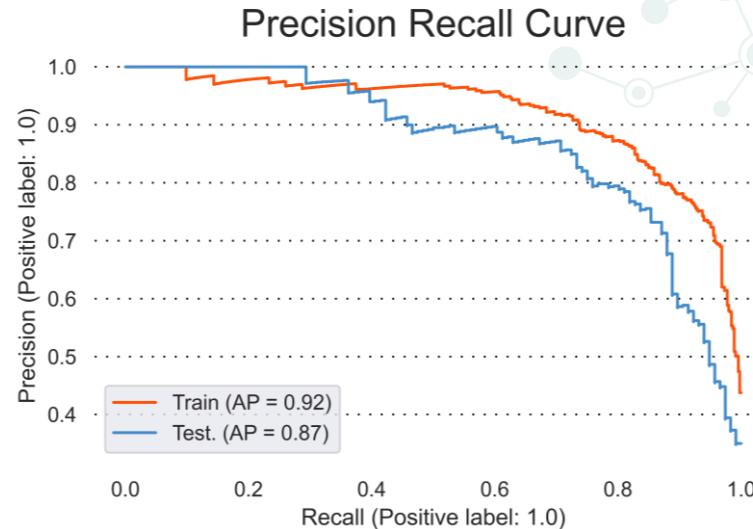
# OVERVIEW

# QSAR MODEL

		Positive	Negative	
				Sensitivity
				$\frac{TP}{(TP + FN)}$
Actual Class	Positive	True Positive (TP)	False Negative (FN) <b>Type II Error</b>	$\frac{TP}{(TP + FN)}$
	Negative	False Positive (FP) <b>Type I Error</b>	True Negative (TN)	
		Precision $\frac{TP}{(TP + FP)}$	Negative Predictive Value $\frac{TN}{(TN + FN)}$	Accuracy $\frac{TP + TN}{(TP + TN + FP + FN)}$

$$F_1 = \frac{2(\text{Precision} \times \text{Recall})}{\text{Precision} + \text{Recall}}$$

## Evaluation metric - Classification



$$AP = \sum_n (R_n - R_{n-1})P_n$$

## OVERVIEW

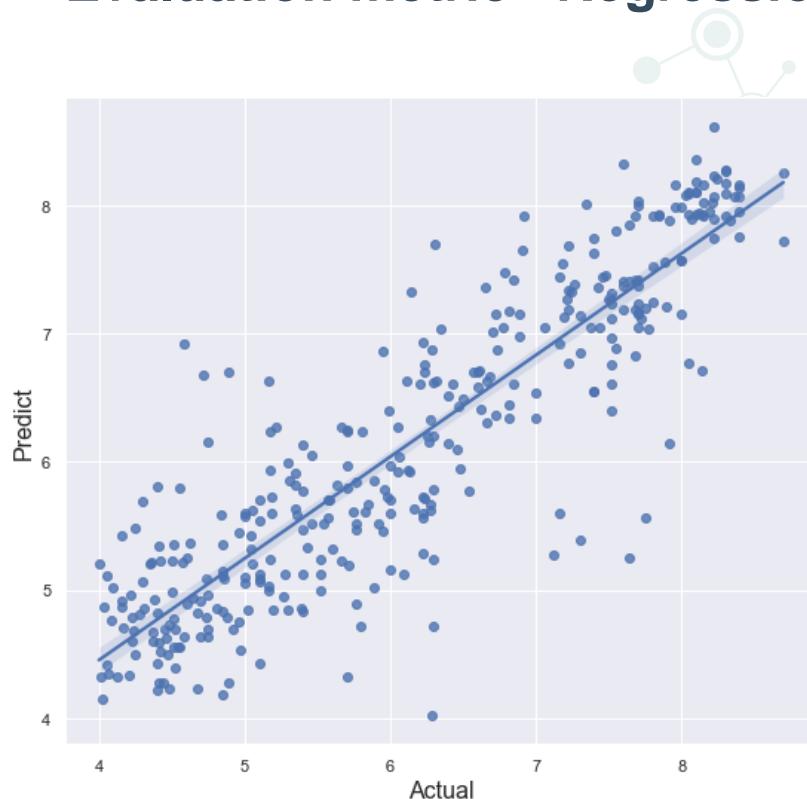
$$R^2 = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y}_i)^2}$$

$$RMSE = \frac{1}{n} \times \sqrt{\|y - \hat{y}\|_2^2}$$

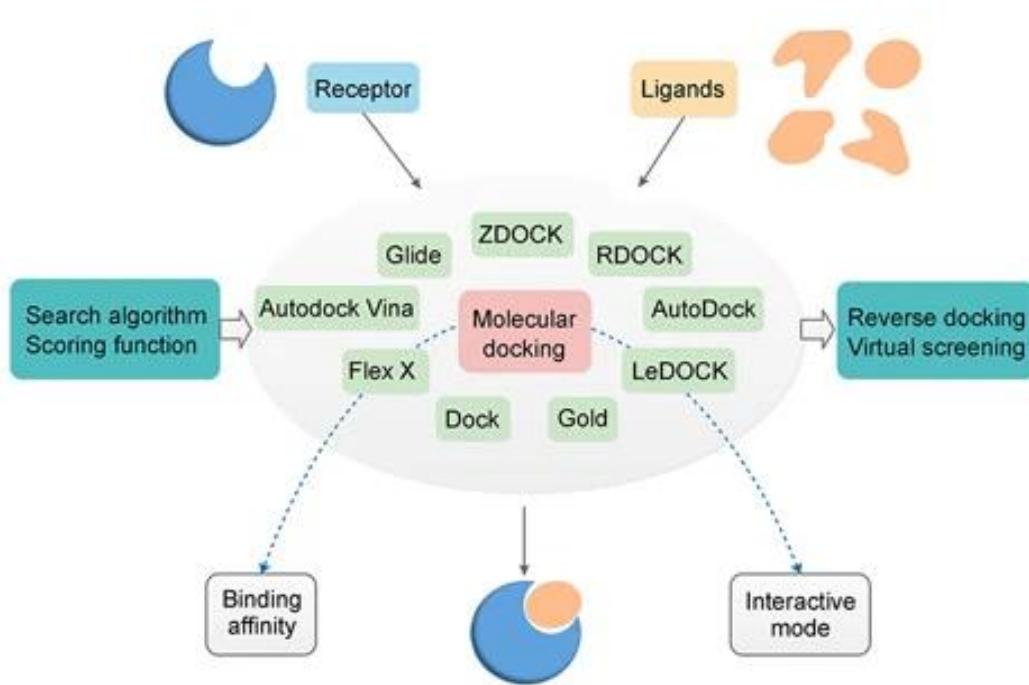
$$MAE = \frac{1}{n} \times \|y - \hat{y}\|_1$$

## QSAR MODEL

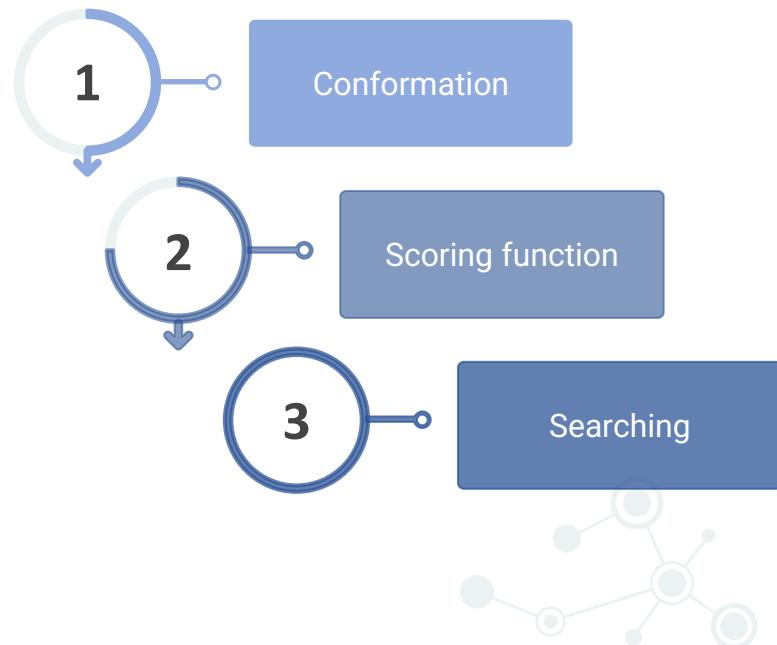
### Evaluation metric - Regression



## OVERVIEW

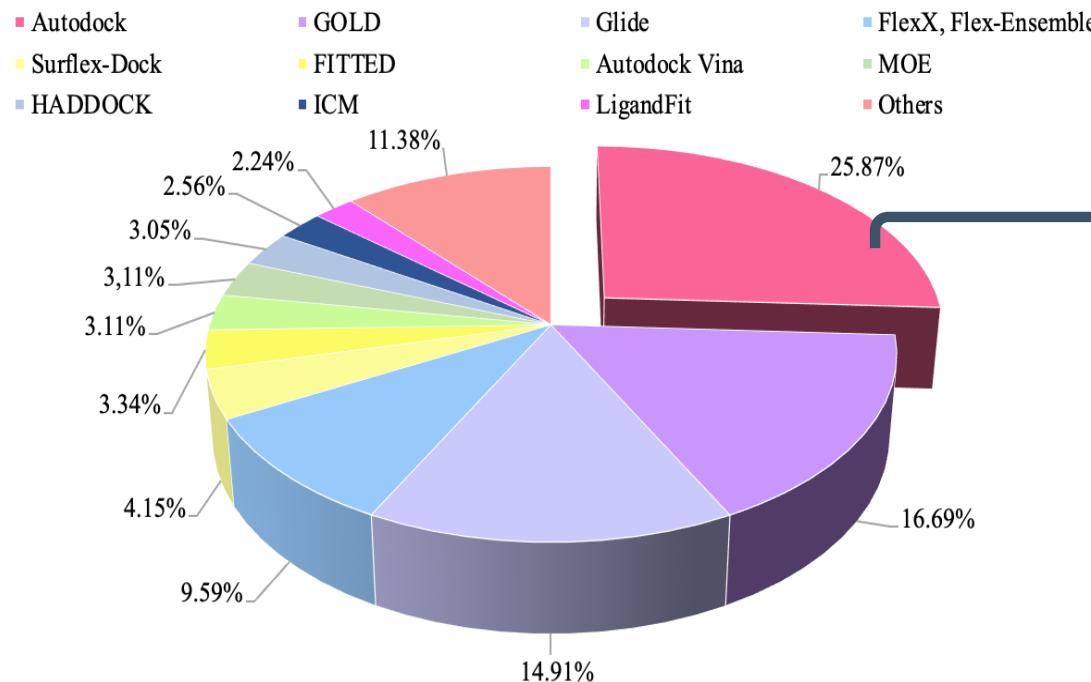


## MOLECULAR DOCKING



## OVERVIEW

## MOLECULAR DOCKING



Autodock had highest citation in ISI Web of Science (2005)



## OVERVIEW

## MOLECULAR DOCKING

1989

**Autodock**

Physics-based scoring functions

01

2010

**Autodock vina 1.1.2**

Empirical Scoring Functions  
Multi-core processing

02

2013

**Smina**

Can modify scoring function

03

**Qvina2**

Increase docking speed by BFGS

04

2021

**Autodock vina 1.2.3**

Python integrated  
Batch-dock  
AD4 scoring function

05

2021

**Autodock-GPU**

GPU Batch-dock

06

2022

**Vina-GPU**

GPU

07



## OVERVIEW

### Active compounds

Active reference for docking model

### Decoy

Inactive reference for docking model

### Metrics

AUC-ROC; G-mean, TPR

### Meaning

Performance of docking softwares

## MOLECULAR DOCKING

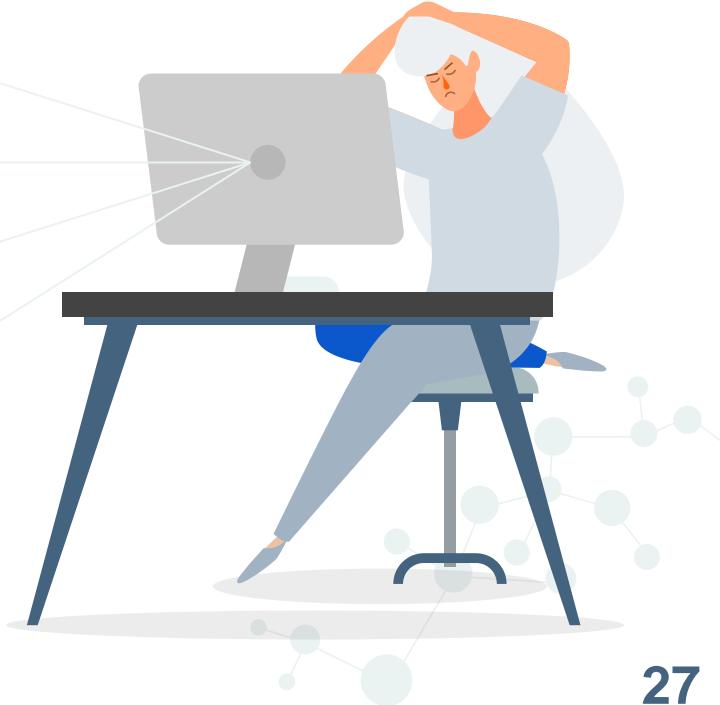
### Retrospective control

01

02

03

04

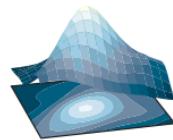


## OVERVIEW

## VIRTUAL SCREENING

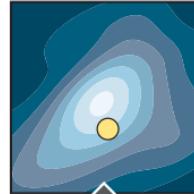
# Computational drug discovery: three schemes

### Functional space



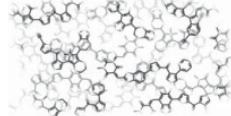
Desired properties (redox potential, solubility, toxicity)

### Simulation



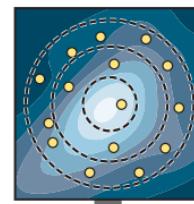
Experiment or simulation (Schrödinger equation)

### Chemical space

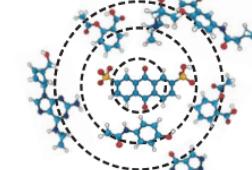


(Drug-like, photovoltaics, polymers, dyes)

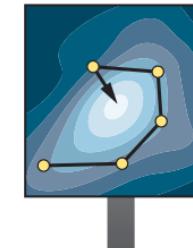
### Virtual screening



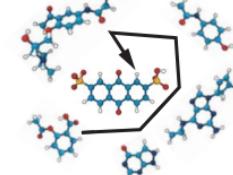
High-throughput virtual screening (e.g., with 3 filtering stages)



### De novo drug design



Optimization, evolutionary strategies, generative models (VAE, GAN, RL)



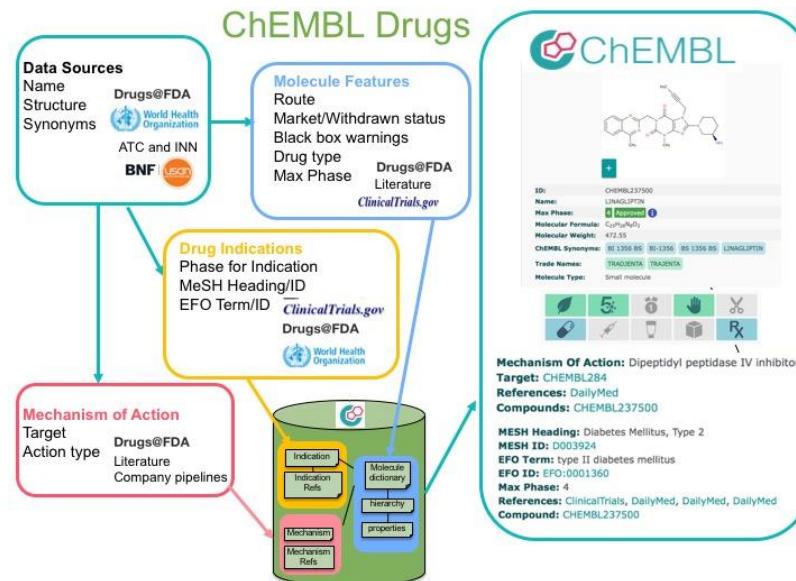


# DATA



# DATA

## DATA FOR DEVELOPING MODELS



## DATA FOR SCREENING

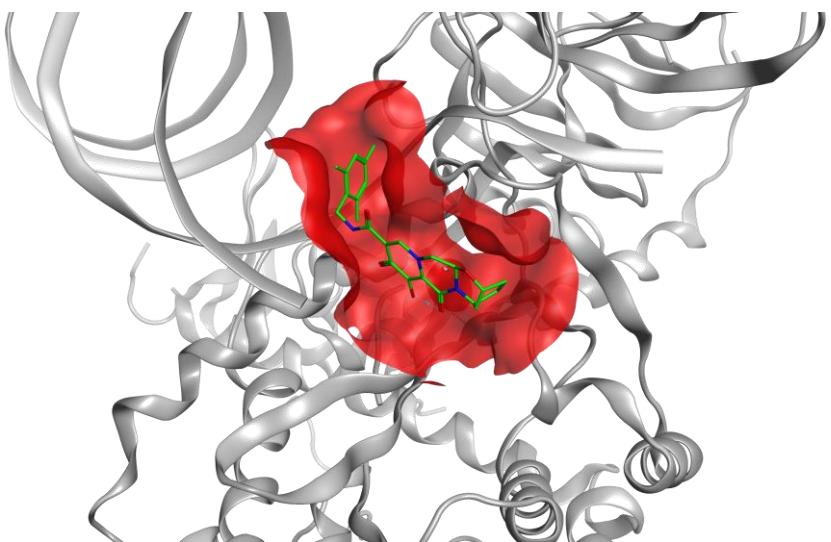


Internal database

2016 compounds



DATA

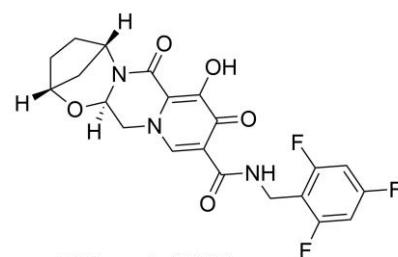


PROTEIN

Cryo-em

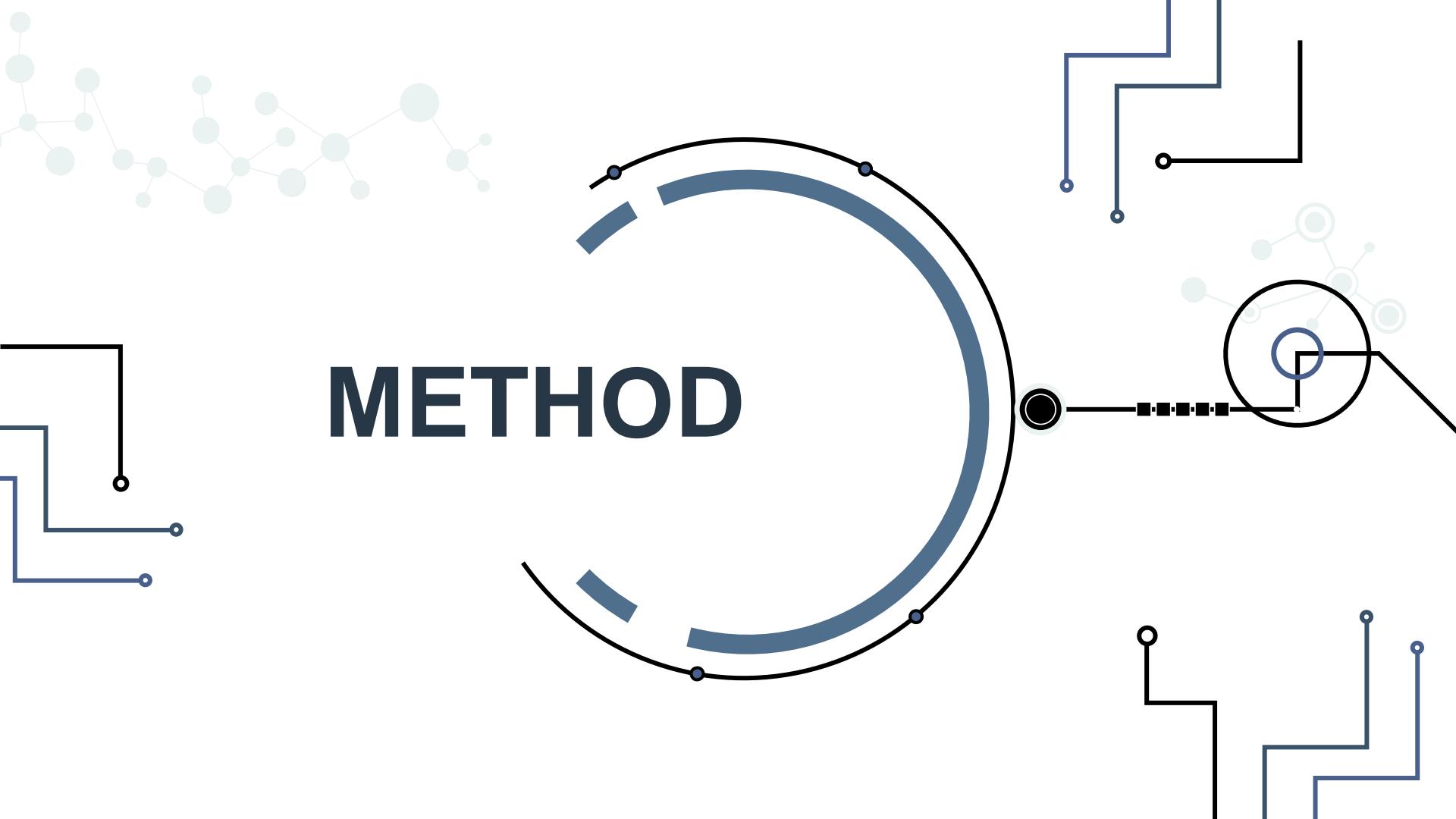
2,9 Å

PDB ID: 6PUW



Bictegravir (2018)

# METHOD



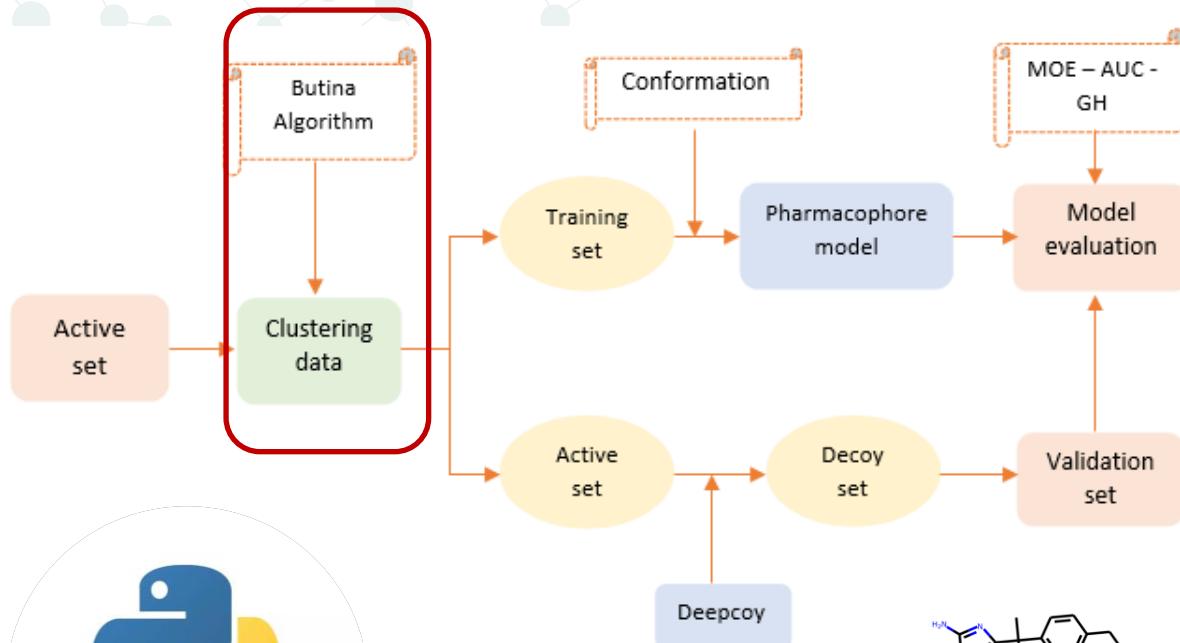


## Method



# PHARMACOPHORE

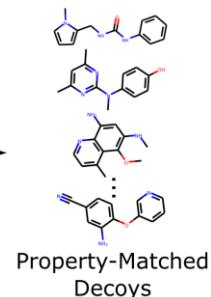
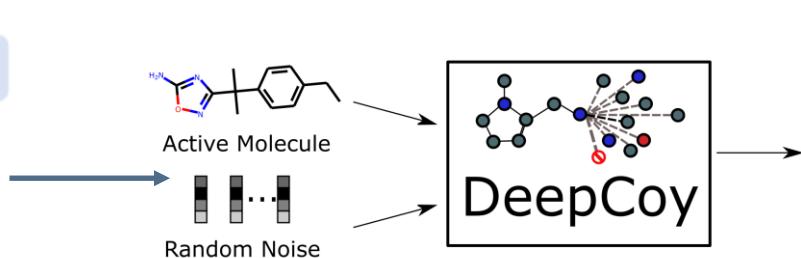
## METHOD



## PHARMACOPHORE



Chemical  
Computing  
Group



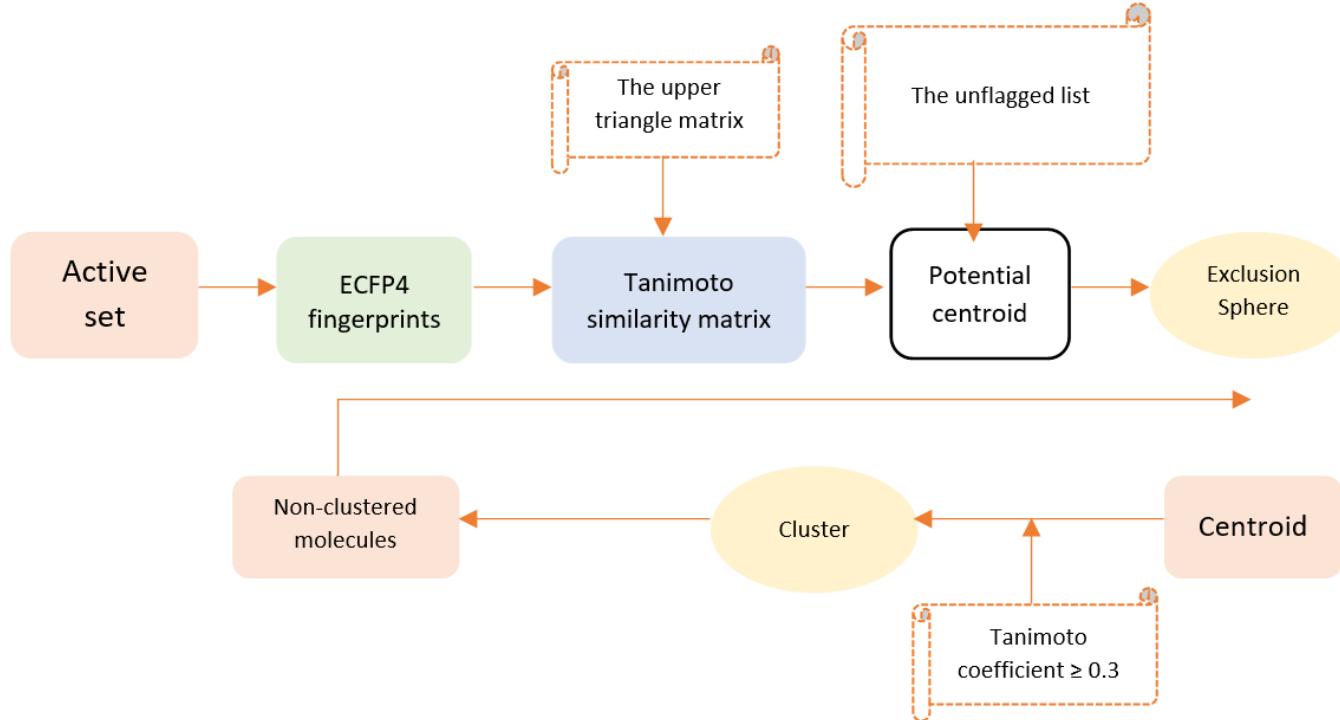
Property-Matched  
Decoys



## METHOD

## PHARMACOPHORE

### Molecules clustering

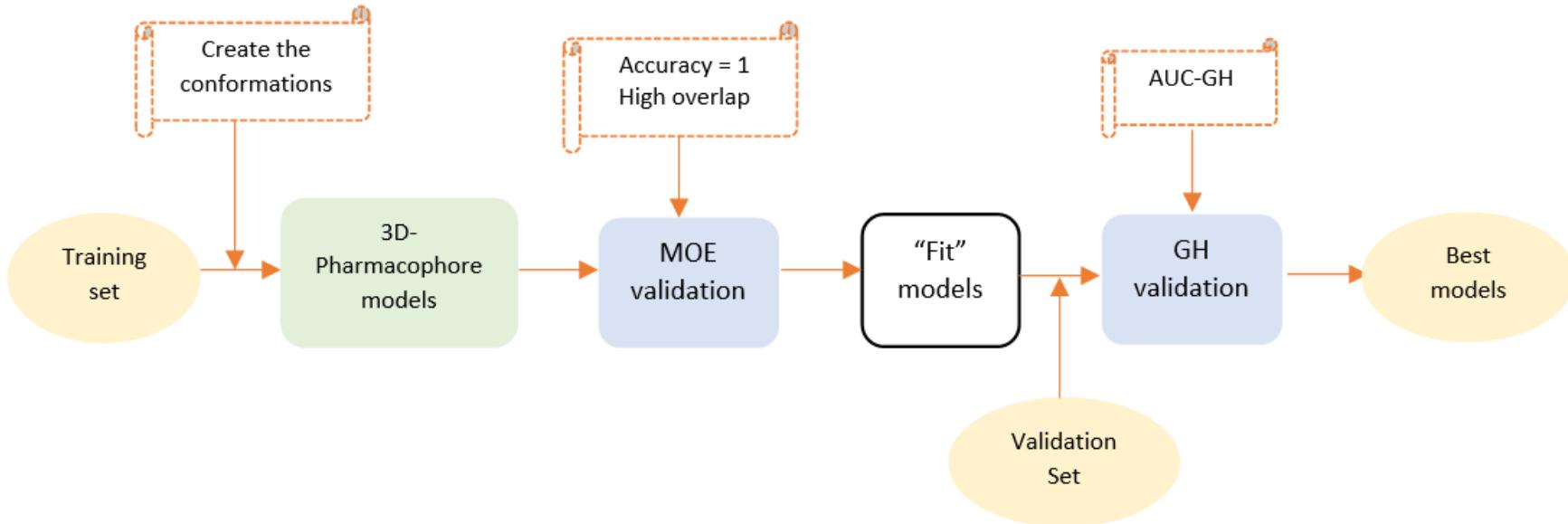




## METHOD

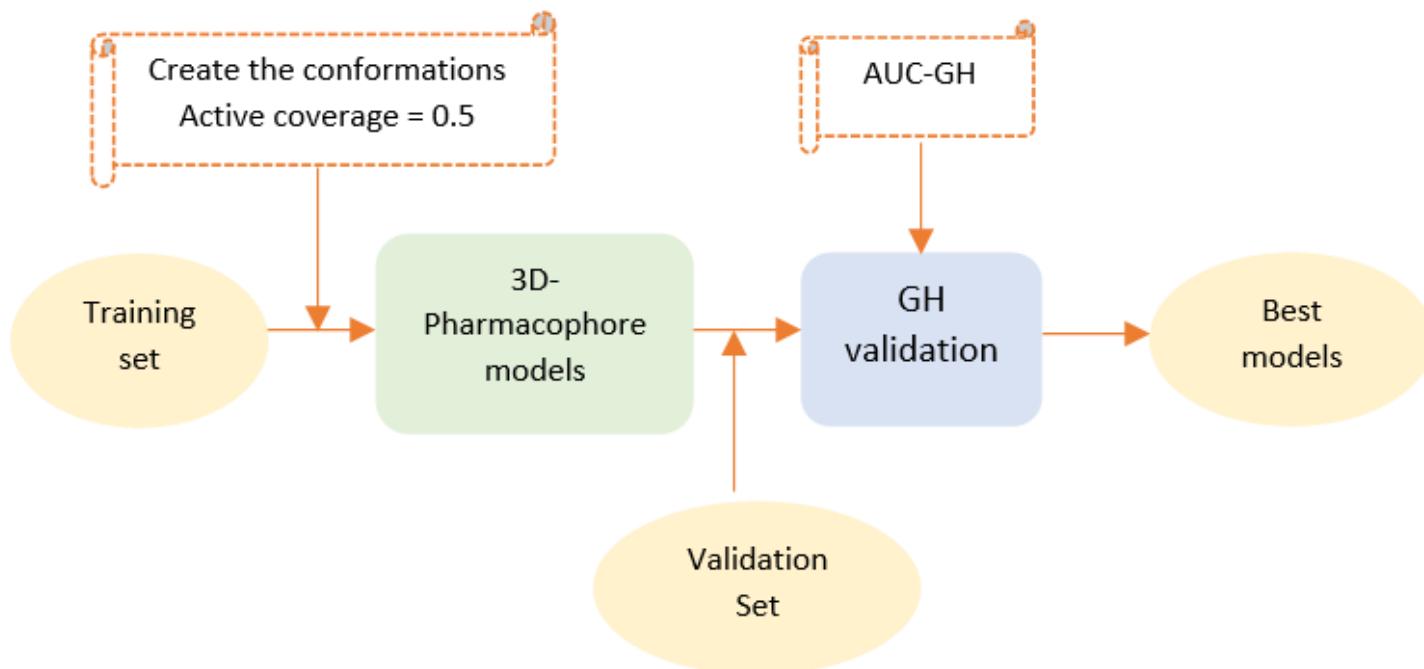
# PHARMACOPHORE

## Local Search





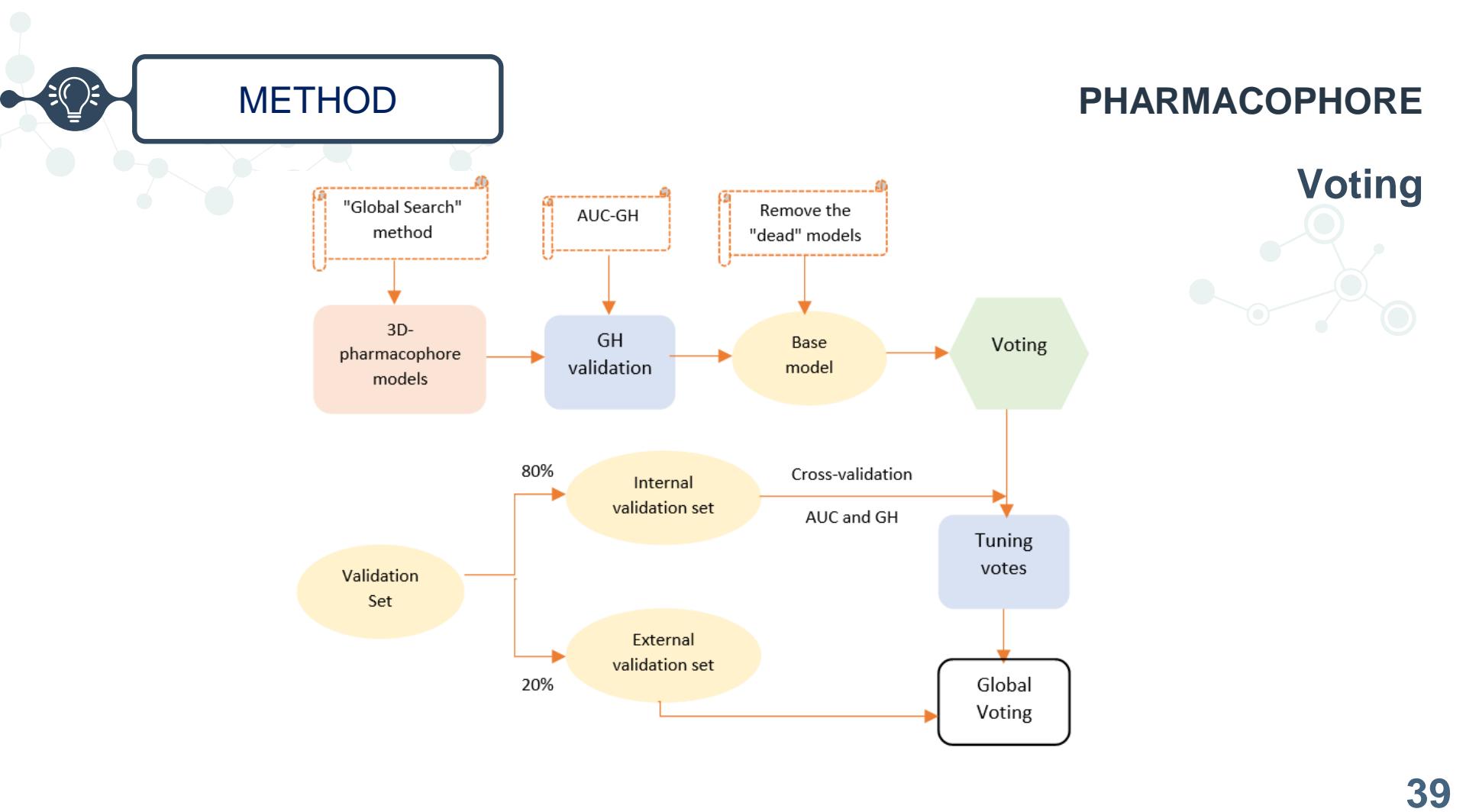
## METHOD



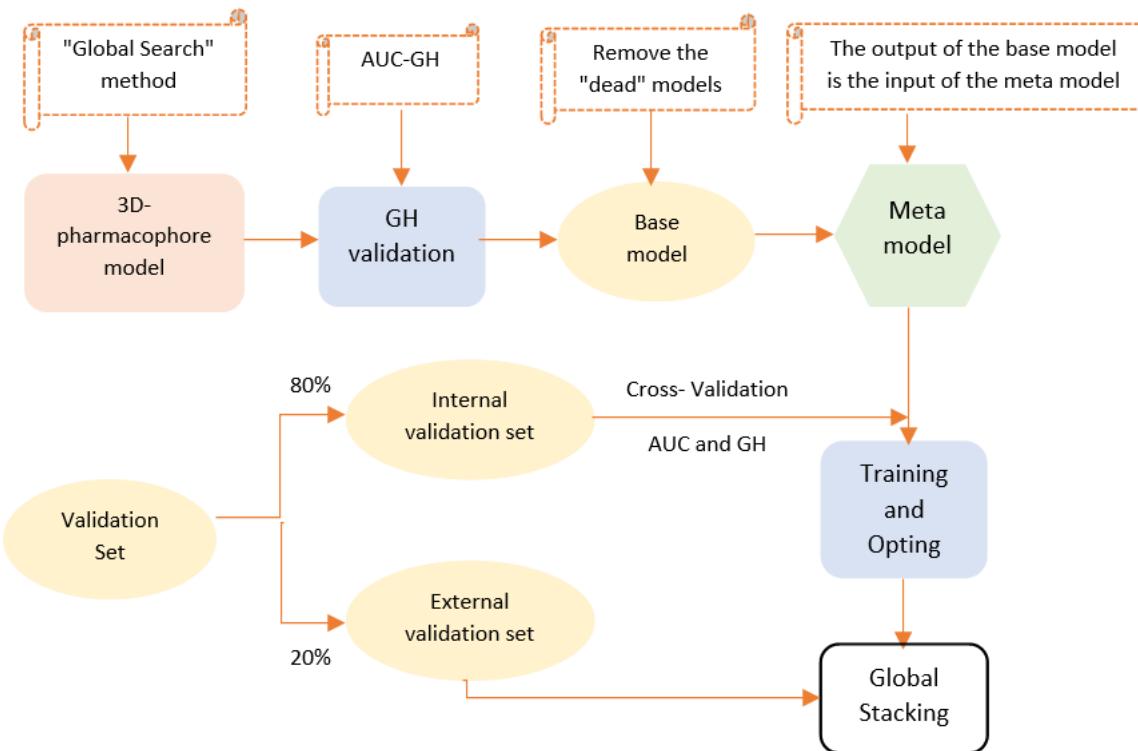
# PHARMACOPHORE

## Global Search





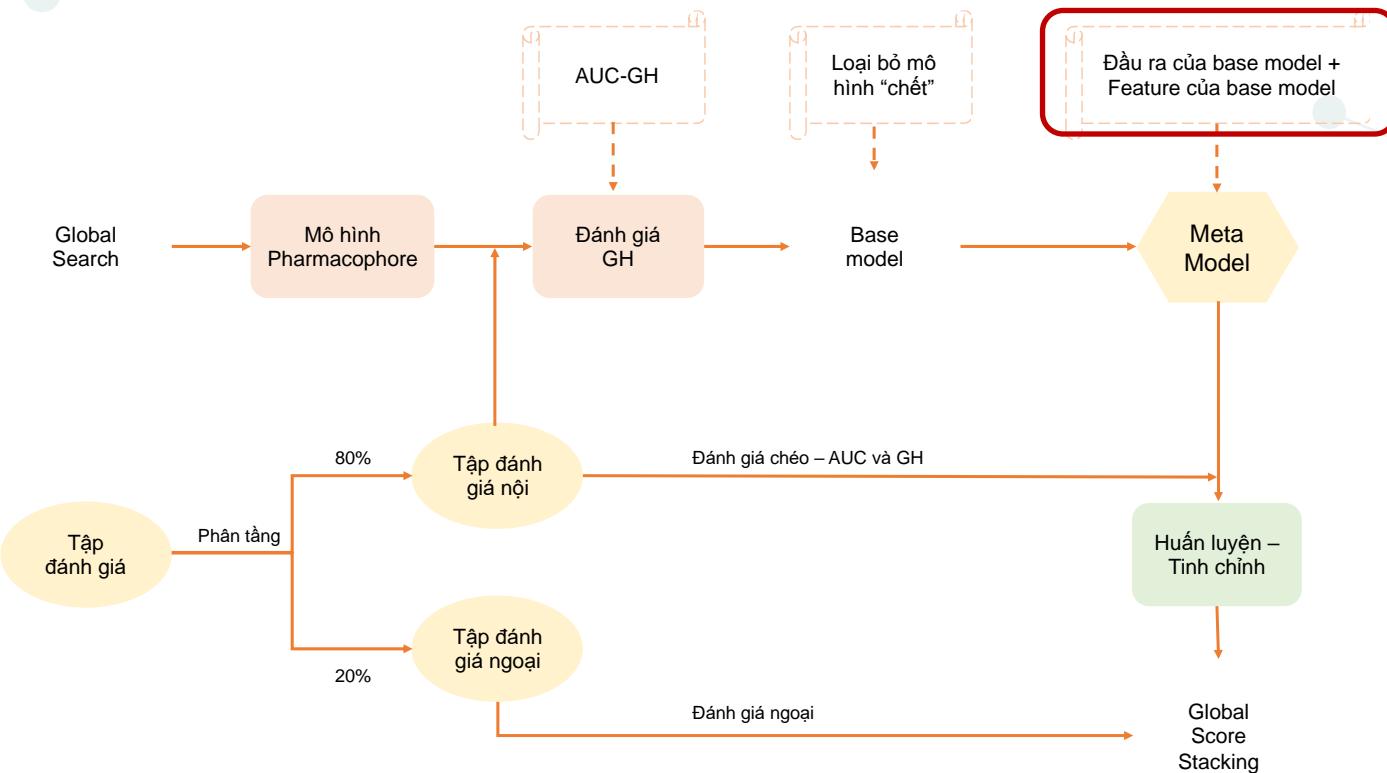
## METHOD



## MÔ HÌNH PHARMACOPHORE

### Stacking

# METHOD



# PHARMACOPHORE

## Score Stacking



## METHOD

### Model

1. Local Search
2. Global Search
3. Voting
4. Stacking
5. Score Stacking
- 6. Feature Score Stacking**
- 7. Optimized Feature Score Stacking**

## PHARMACOPHORE

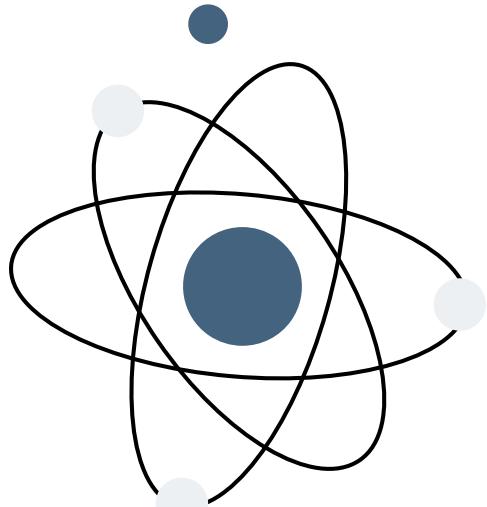
### EVALUATION

#### Evaluation method

- **Cross validation:** post hoc Wilcoxon
- **External validation:** generalization
- **Metric:**  
AUC > GH > F1

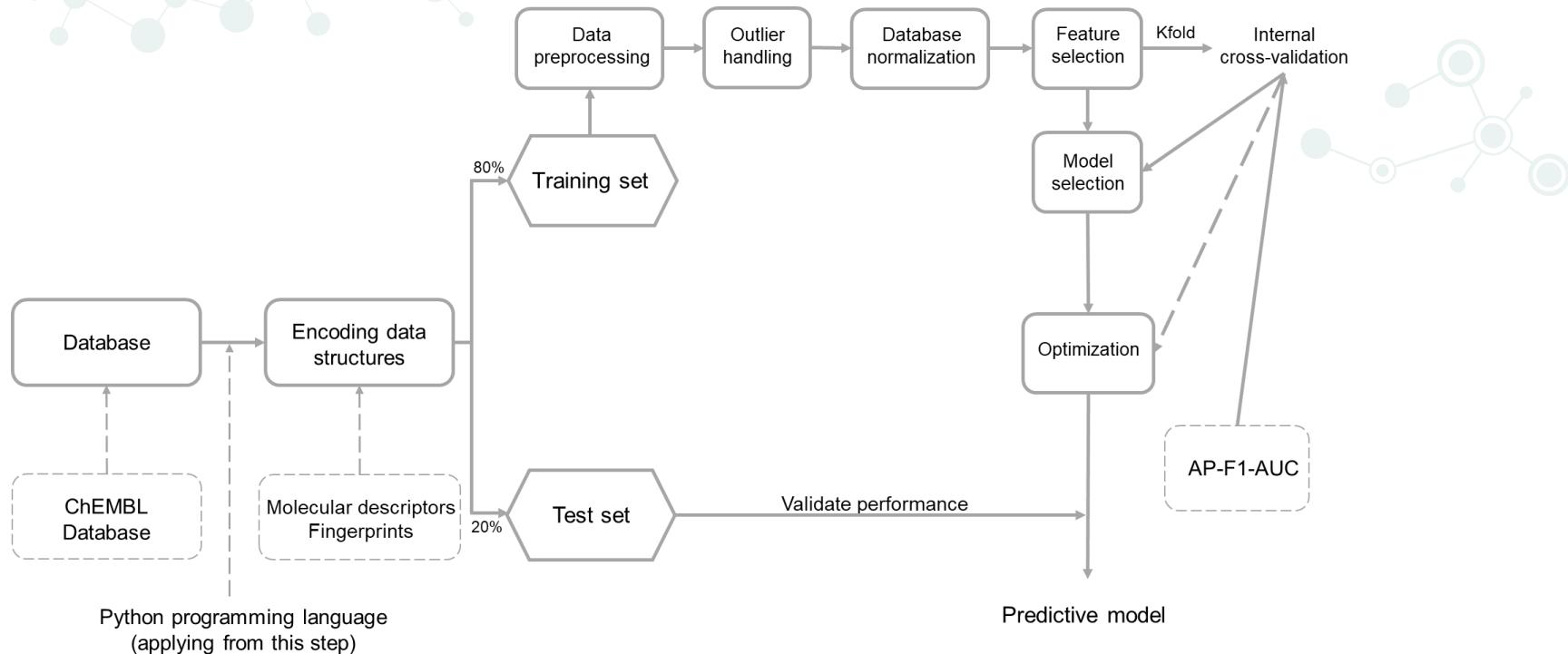


## METHOD



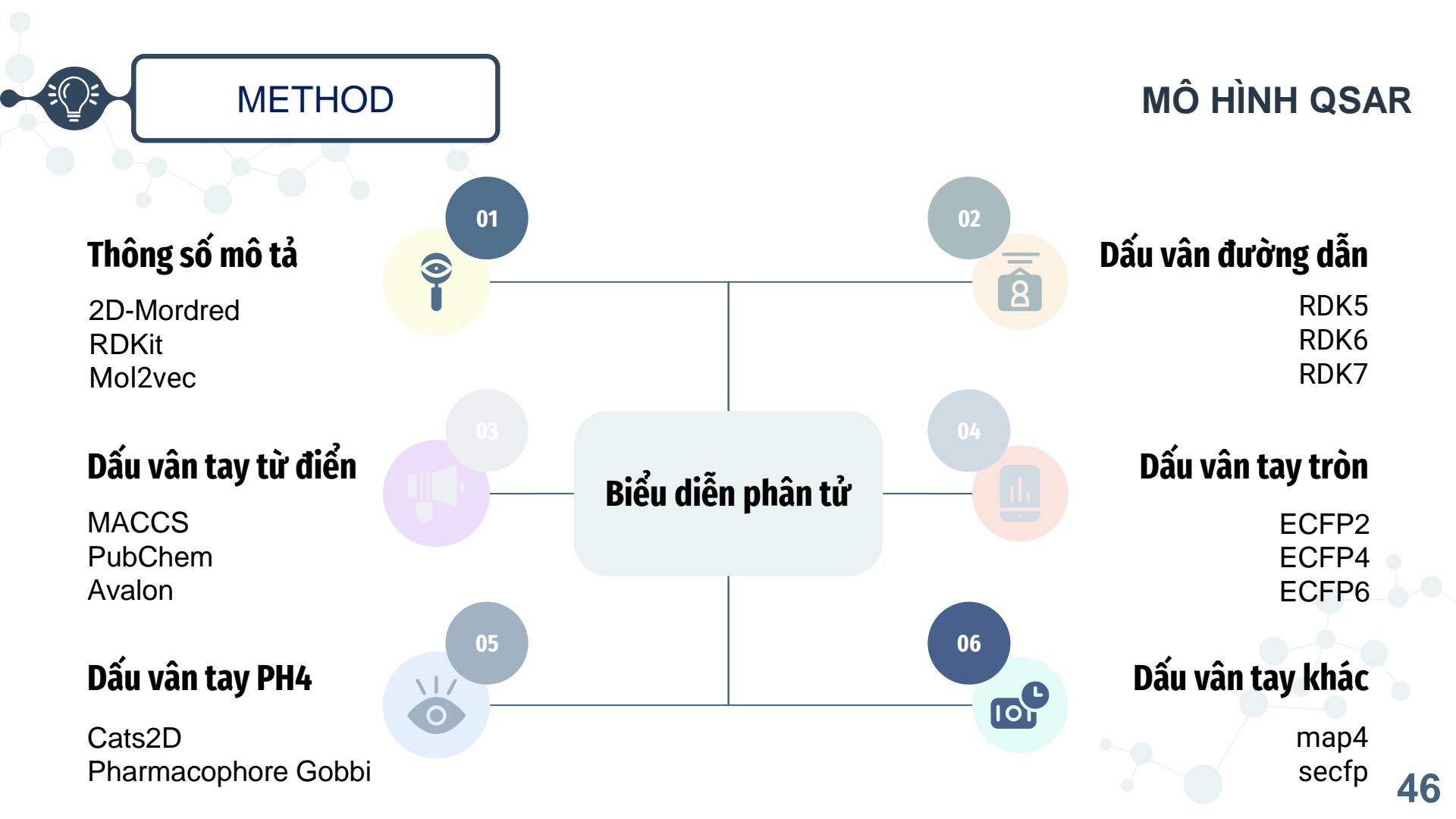
# QSAR Model

# MÔ HÌNH QSAR



Source code

# MÔ HÌNH QSAR





QSAR

Data-centric



**01**

Xây dựng  
cơ sở dữ liệu



**02**

Khai phá  
dữ liệu



**03**

Lựa chọn bộ  
dữ liệu tối ưu



**04**

Lựa chọn  
đặc trưng



**05**

Lựa chọn  
mô hình



**06**

Tinh chỉnh  
mô hình



## PHƯƠNG PHÁP



# MOLECULAR DOCKING

03



## METHOD

Protein Data Bank

Cấu trúc protein

Thông số  
khoang gắn kết

Google Patents  
Tài liệu khoa học,...

Tập hoạt tính  
(bao gồm ligand  
đồng kết tinh)

Tập decoy

Retrospective  
Control

Tối ưu hoá  
mô hình

Gắn kết  
phân tử

Ligand nghiên cứu

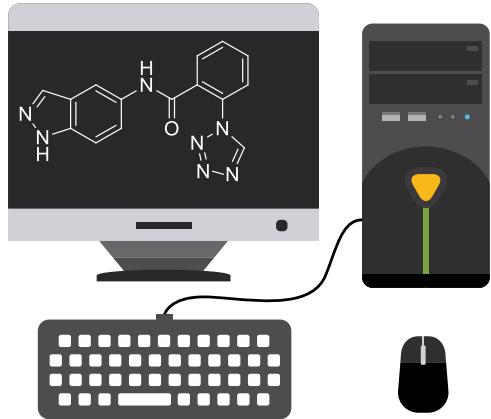
Thư viện nội bộ

Chất tiềm  
năng

## MOLECULAR DOCKING



## METHOD

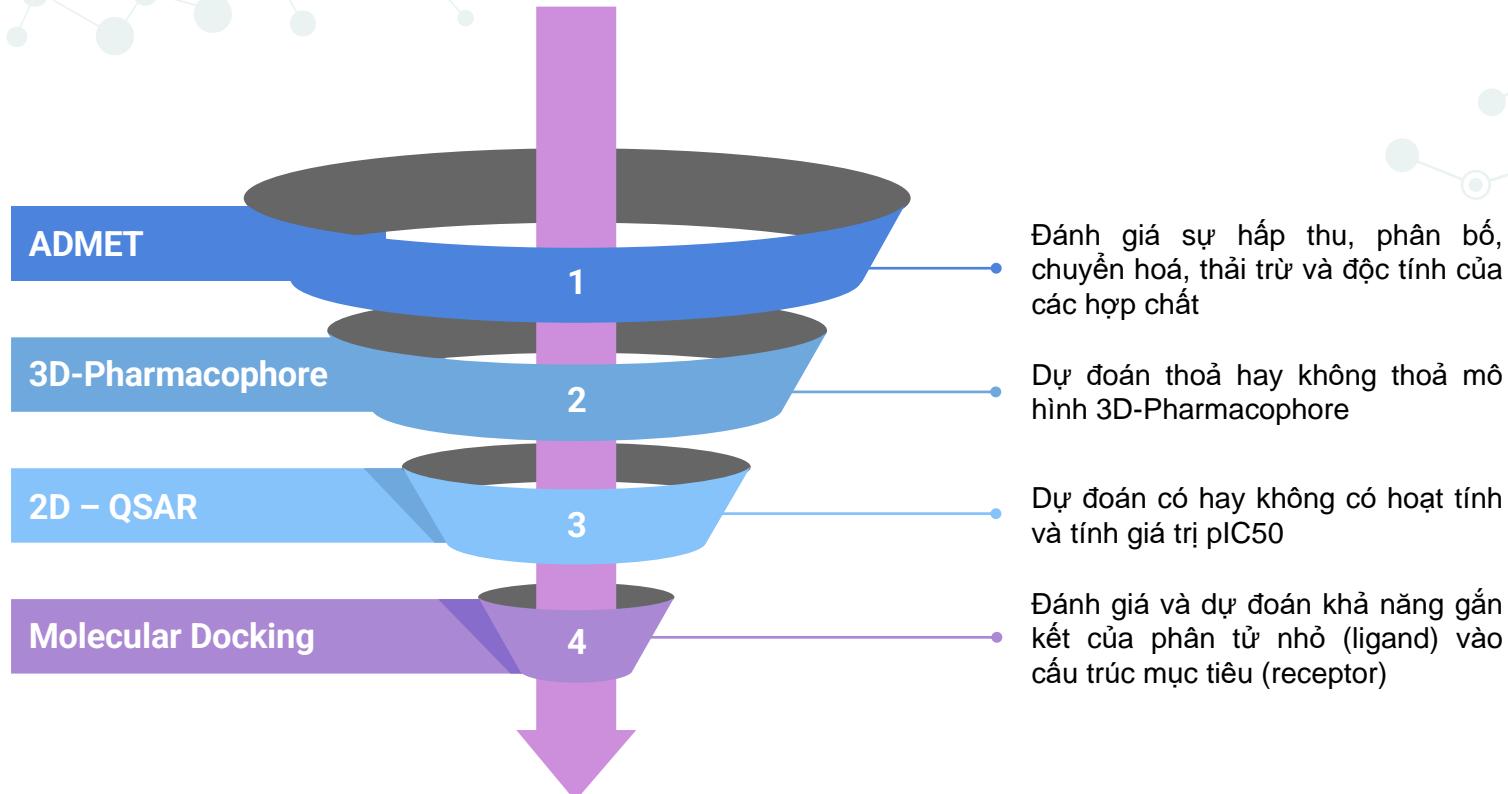


# VIRTUAL SCREENING

04



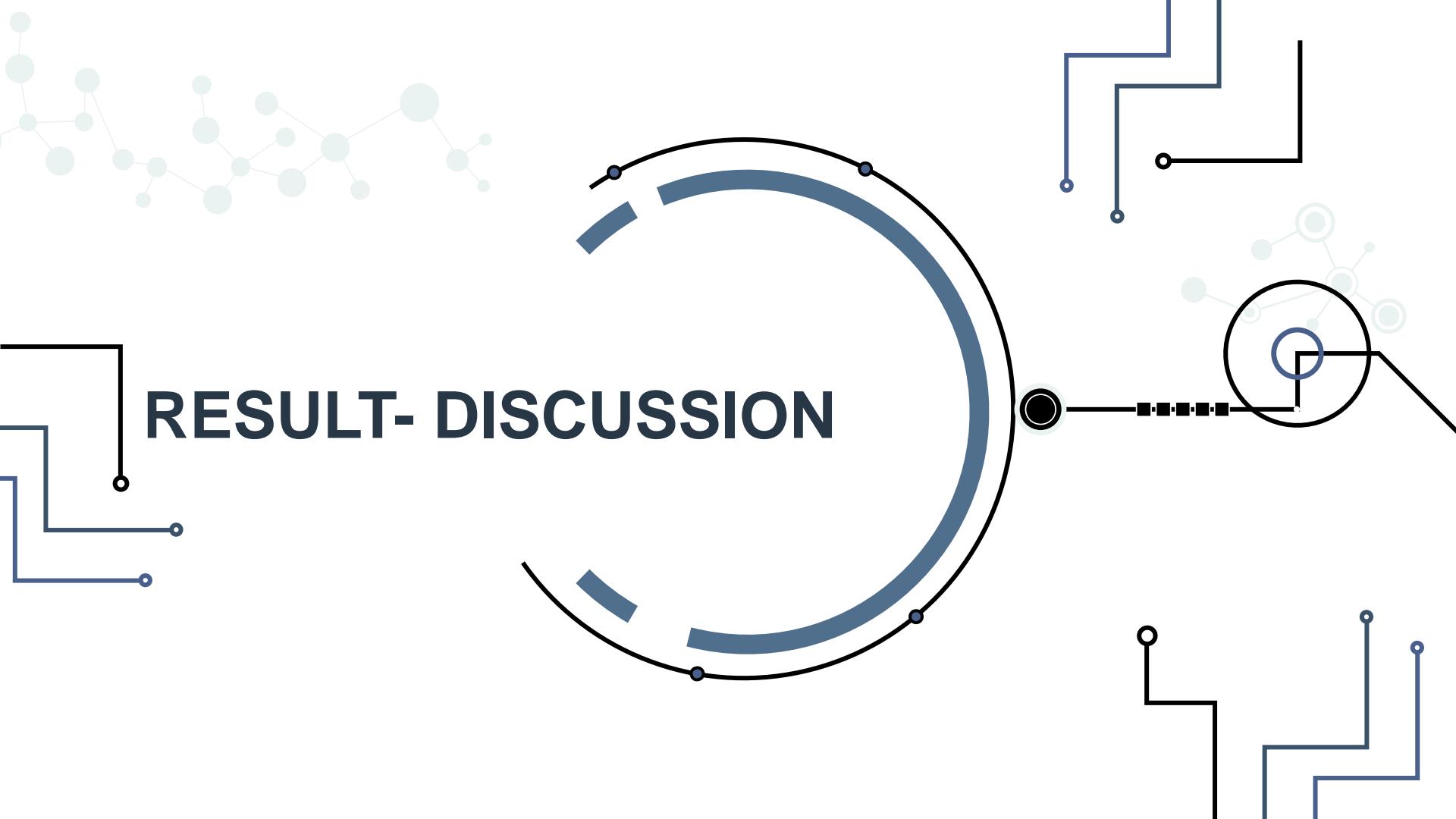
## METHOD



## VIRTUAL SCREENING



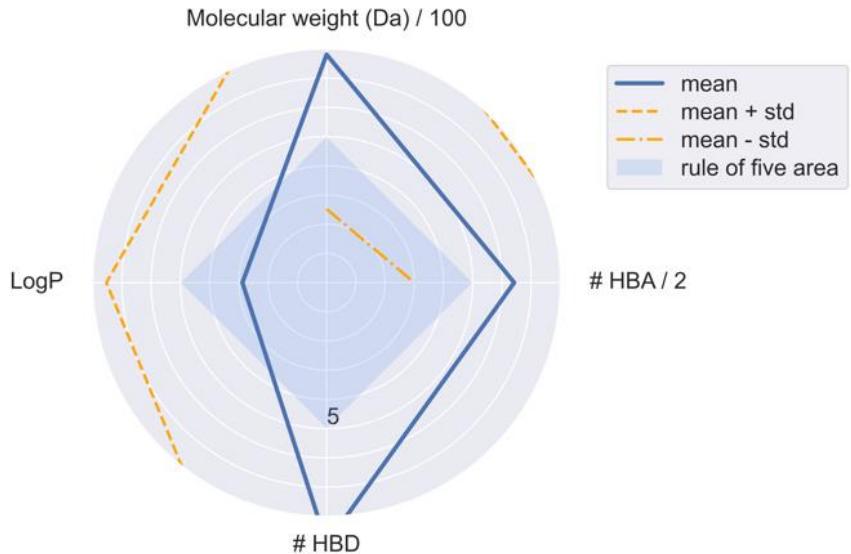
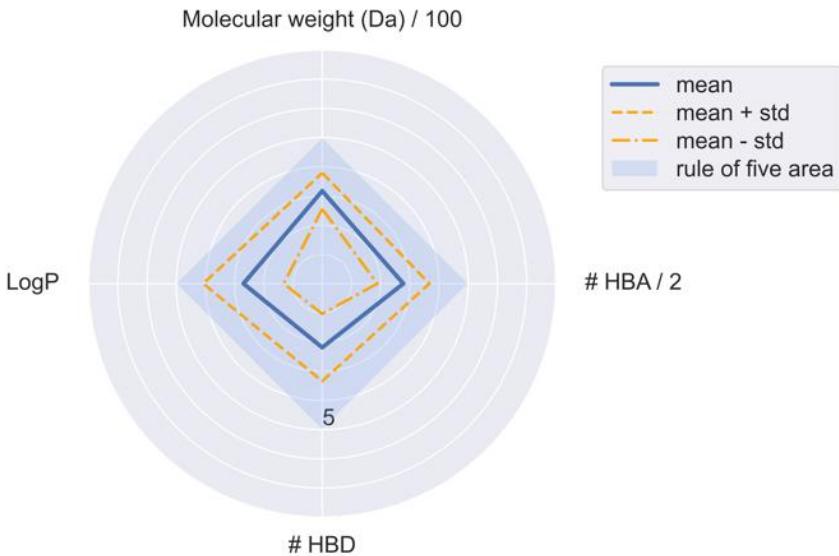
# RESULT- DISCUSSION





# RESULT

2445  
compounds

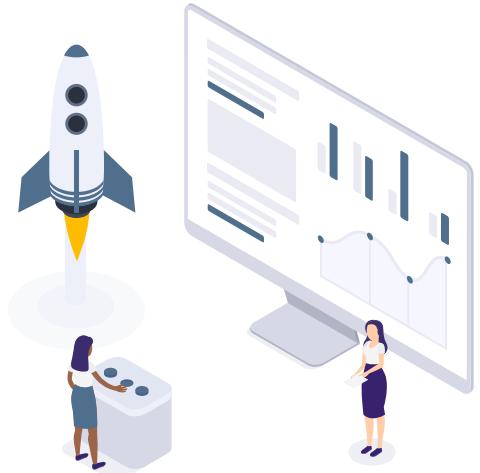


1

RESULT

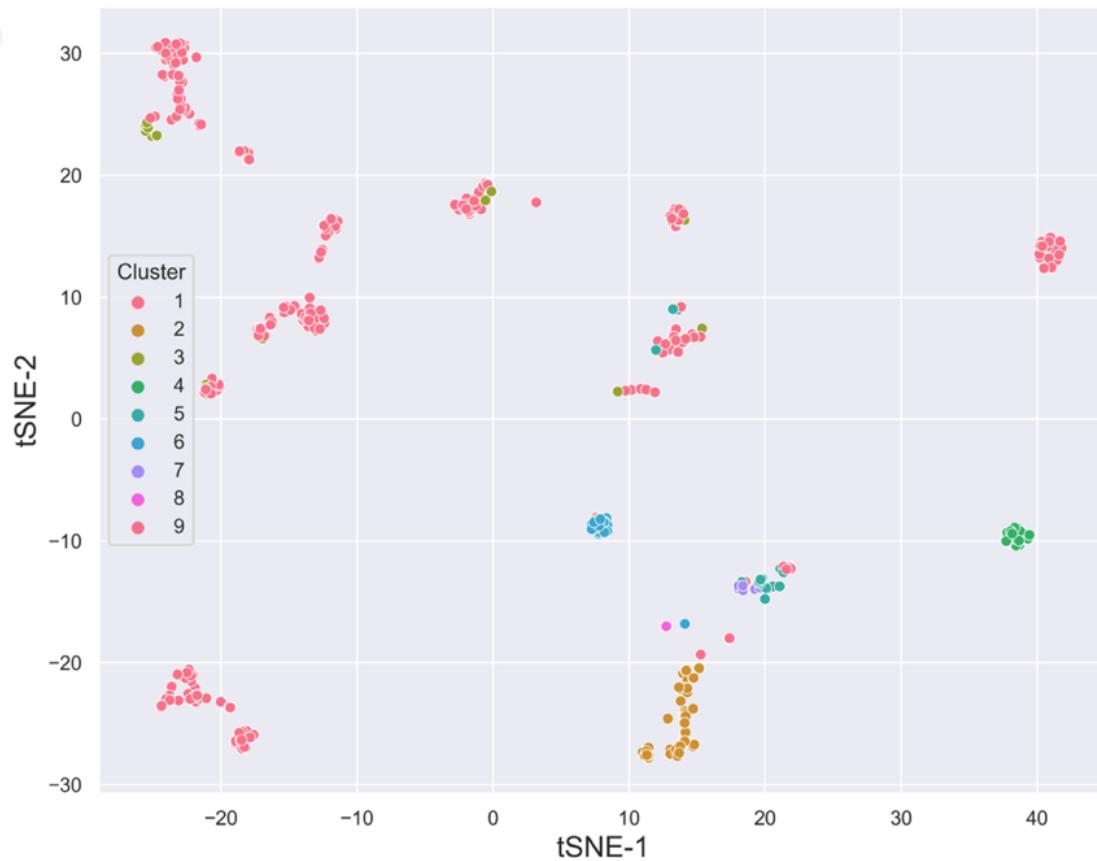
# Pharmacophore

Ligand-based pharmacophore





## RESULT



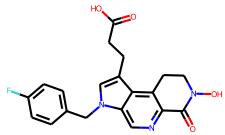
Pharmacophore

Clustering

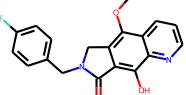




CHEMBL1914556



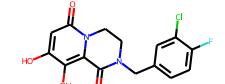
CHEMBL584360



CHEMBL4126686



CHEMBL1773405



CHEMBL209440



CHEMBL385951



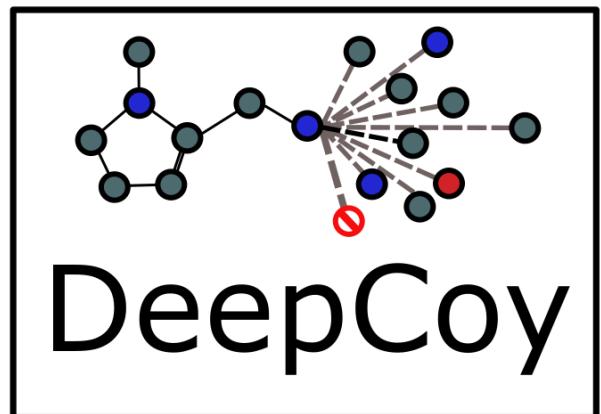
CHEMBL237727

CHEMBL4463247

CHEMBL429327

# Pharmacophore





Metric	Values
AUC-ROC 1NN	0,533
AUC-ROC RF	0,702
DOE score	0,065
Doppelganger score mean	0,271
Doppelganger score max	0,421

Pharmacophore

Decoy





Pharmacophore

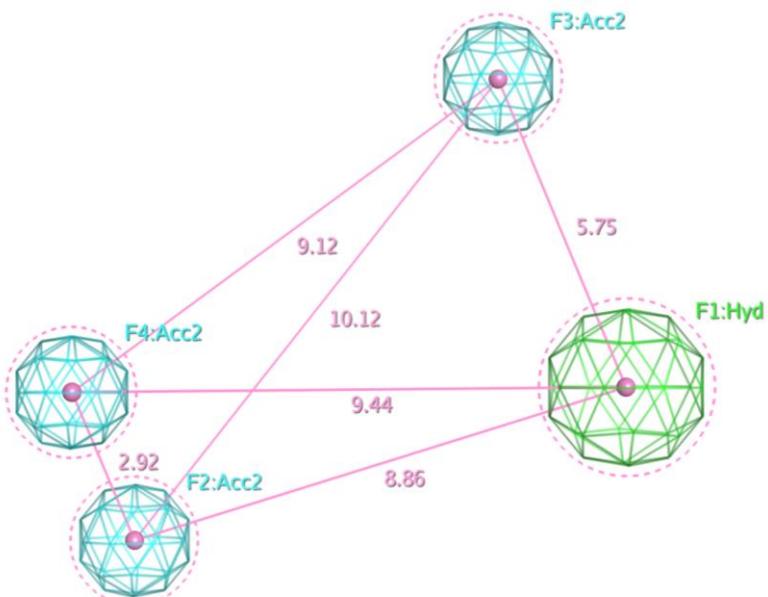
Local search

Model	S <sub>e</sub>	S <sub>p</sub>	P <sub>r</sub>	AUC	GH
Haaa_7	0,814	0,573	0,037	0,794	0,133
Haaa_3	0,765	0,482	0,029	0,632	0,103
Haaa_4	0,8	0,6	0,039	0,748	0,138
Haaa_6	0,403	0,631	0,021	0,489	0,074
....					



## RESULT

### Haaa\_7 model



Pharmacophore

Local search





## RESULT

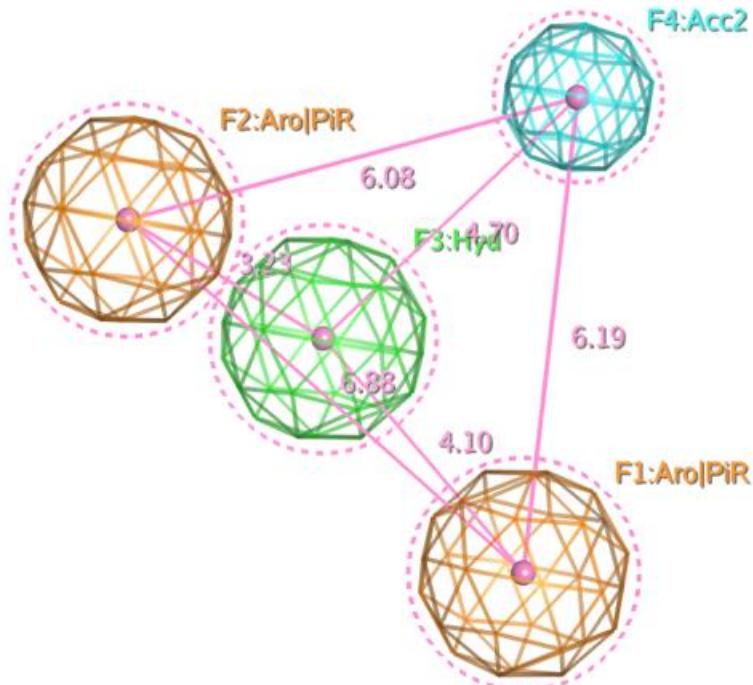
Pharmacophore  
Global Search

12 models have AUC > 0,7

Model	S <sub>e</sub>	S <sub>p</sub>	P <sub>r</sub>	AUC	GH
RRHa_4	0,834	0,894	0,137	0,87	0,278
Haaa_7	0,814	0,573	0,037	0,794	0,133
RRHa_1	0,761	0,812	0,075	0,791	0,200
RHaa_7	0,846	0,546	0,036	0,775	0,130
....					



## RRHa\_4 model



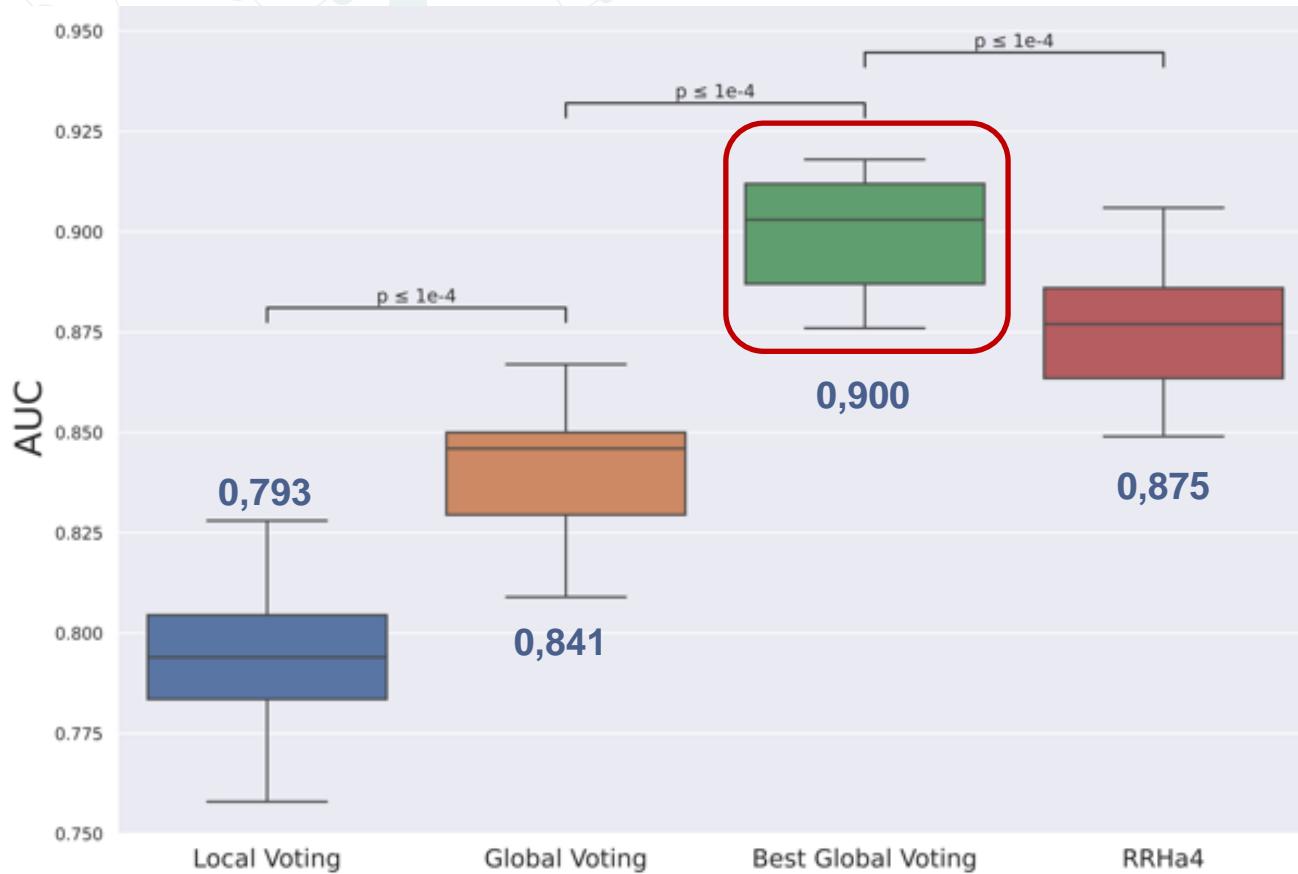
Pharmacophore  
Global search



- Cover 7
- F1 and F2 – Aromatic (Aro:  $r= 1,4\text{\AA}$ )
- F3 is hydrophobic (Hyd: $r=1,4\text{\AA}$ )
- F4 hydro acceptor (Acc2:  $r= 1,0\text{\AA}$ )



# RESULT



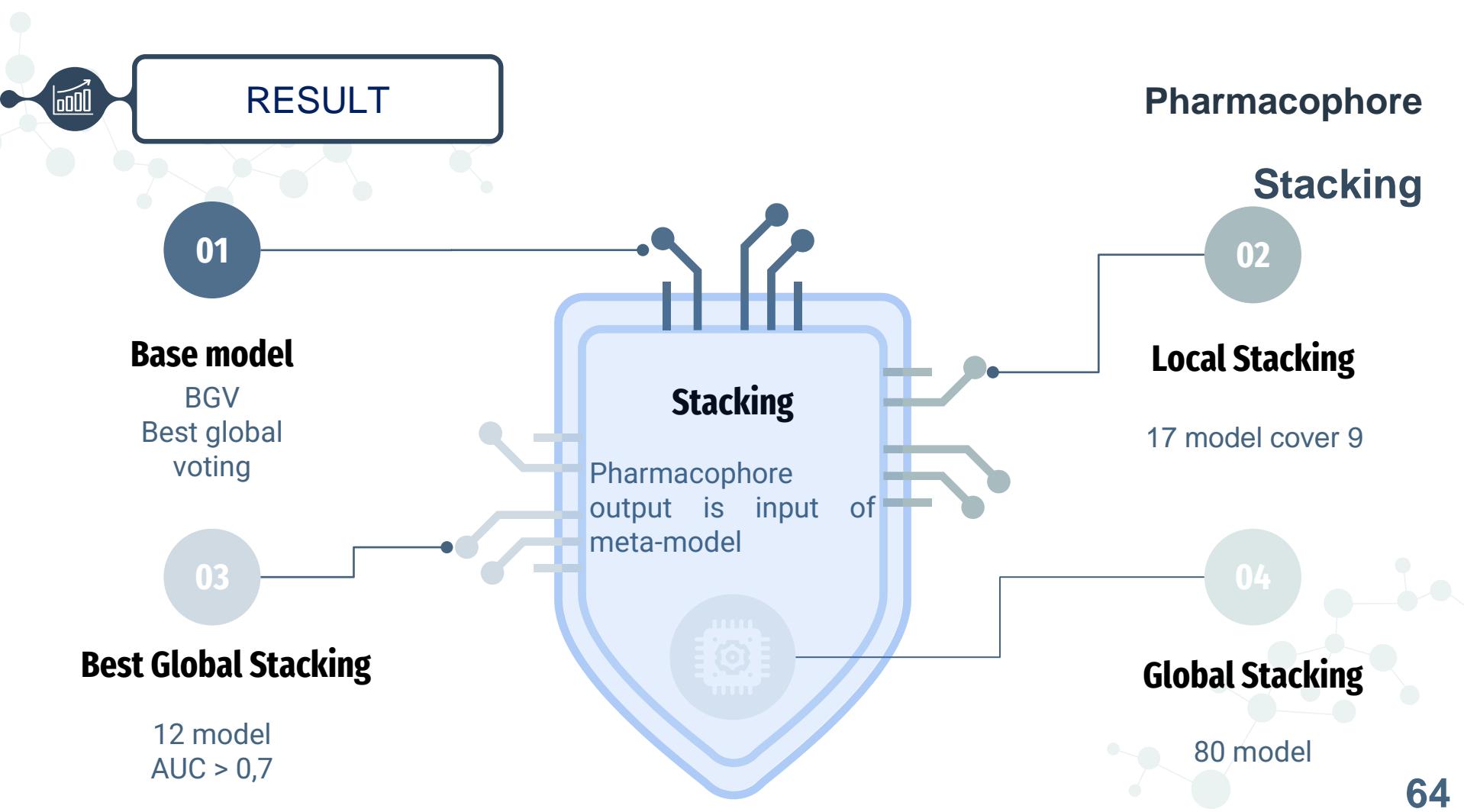
Pharmacophore

Voting

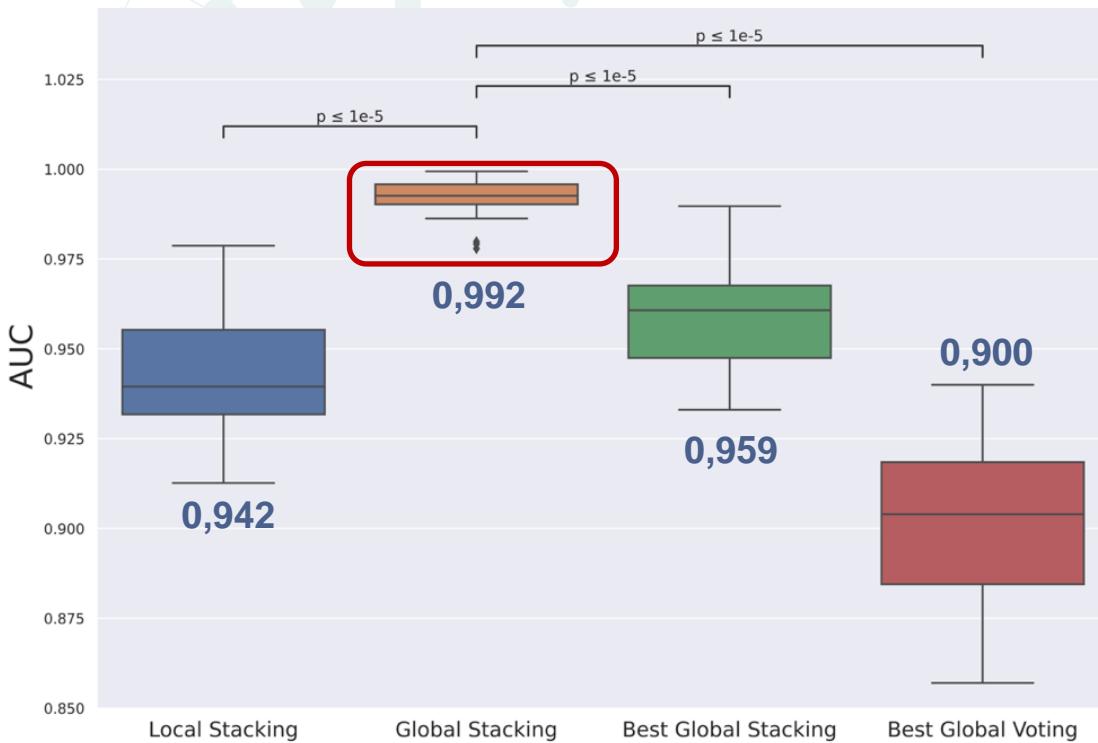


*Best global voting:*

- $\text{AUC} = 0,900 \pm 0,015$
- $\text{GH} = 0,334 \pm 0,049$



# RESULT



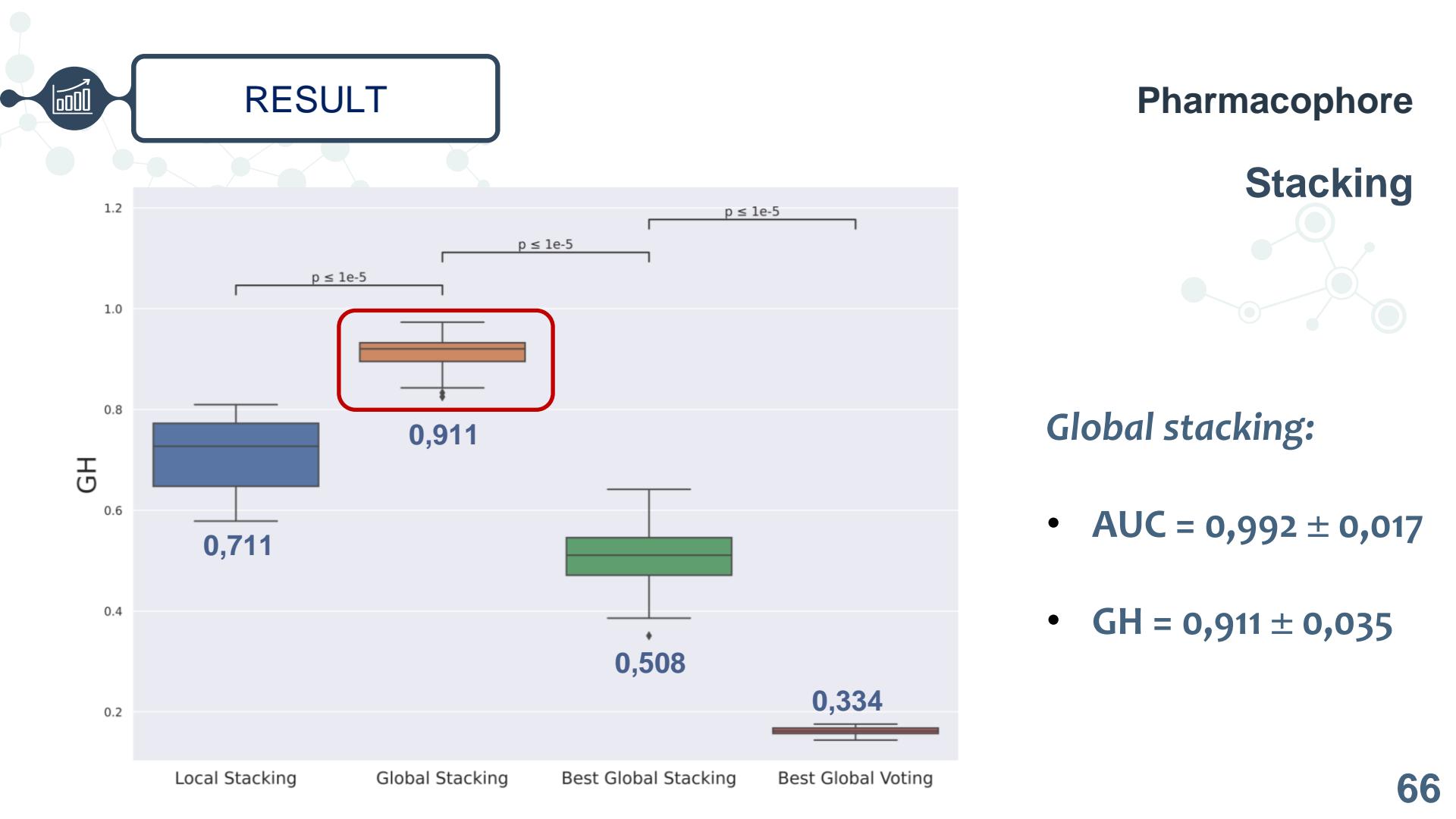
Pharmacophore

Stacking



*Global stacking:*

- $\text{AUC} = 0,992 \pm 0,017$
- $\text{GH} = 0,911 \pm 0,035$





01



Enhance data

**GSS model**  
Using rescore from  
pharmacophore model

02



Feature selection

**F-GSS Model**  
GSS feature selection  
XGBoost: n\_estimators =  
300 max\_depth = 4

Pharmacophore  
Score Stacking

03



Model optimization

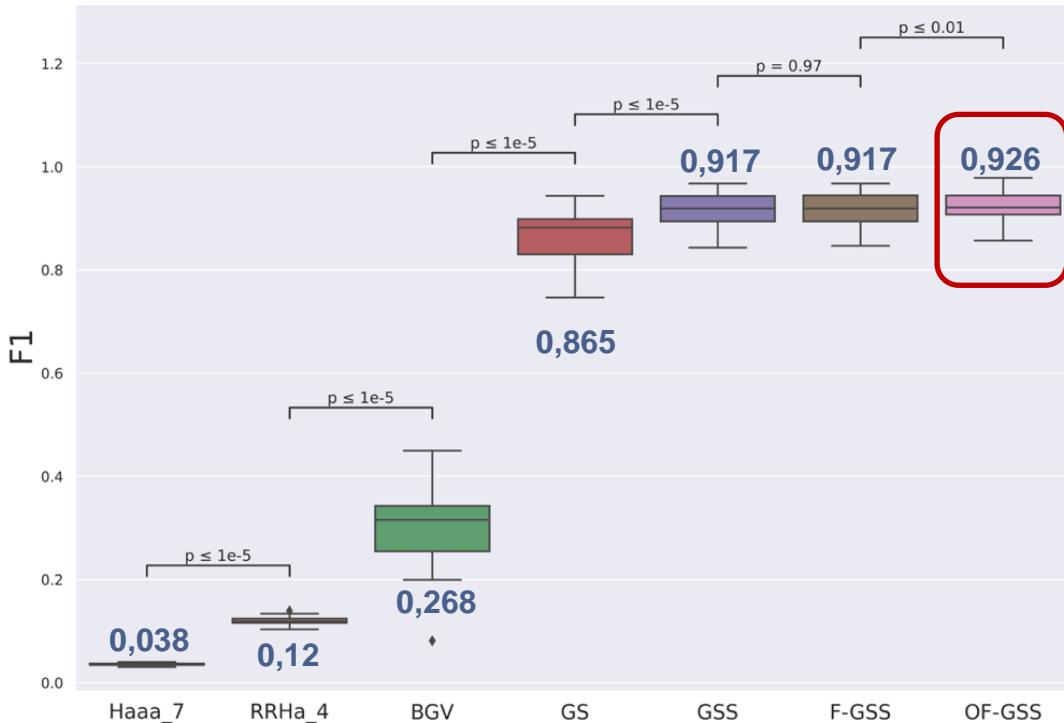
**OF-GSS Model**  
Optimize F-GSS



## DISCUSSION

## Pharmacophore

### Performance comparison



1. Local search: **Haaa\_7**
2. Global search: **Rhaa**
3. **BGV**: 12 model with AUC > 0,7
4. **GS**: stacking 80 model
5. **GSS**: using rescore
6. **F-GSS**: feature selection of GSS
7. **OF-GSS**: optimization of F-GSS

## DISCUSSION



Pharmacophore

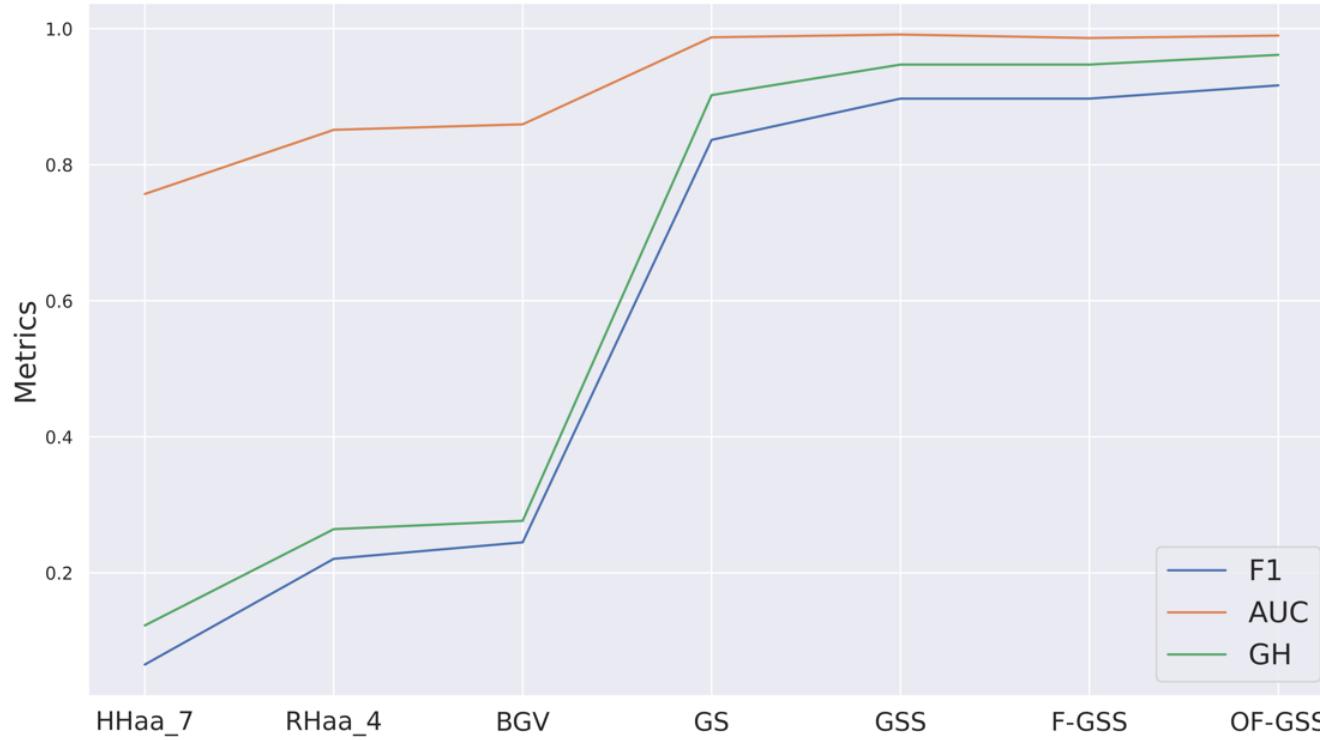
## Performance comparison





## Pharmacophore

## External Validation



### OF-GSS

- F1 = 0,917
- AUC = 0,990
- GH = 0,962



RESULT

2 QSAR



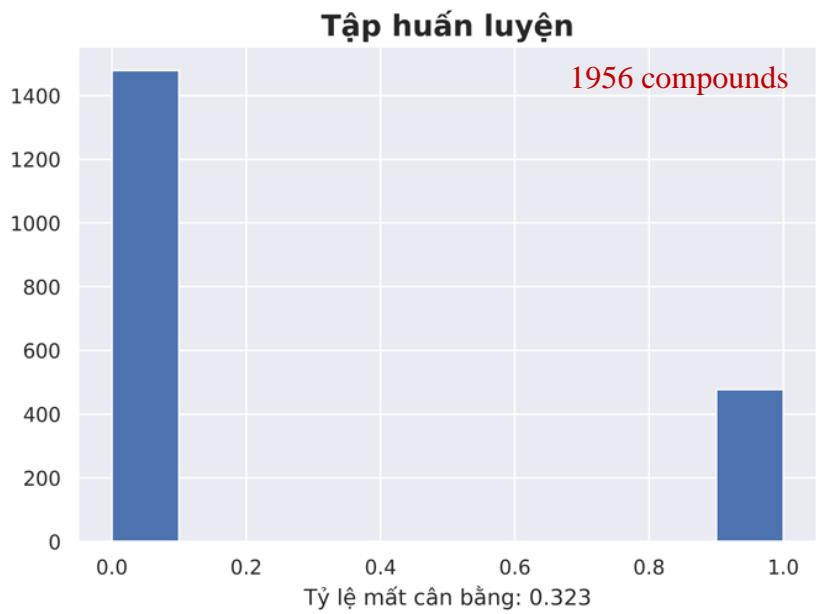


## 2.1. Mô hình học máy phân loại

---

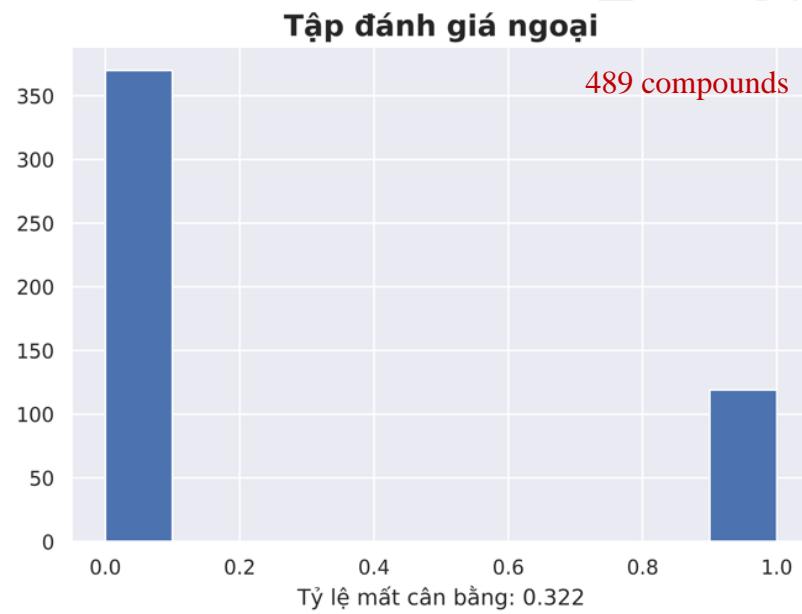


## RESULT



## QSAR – Classification

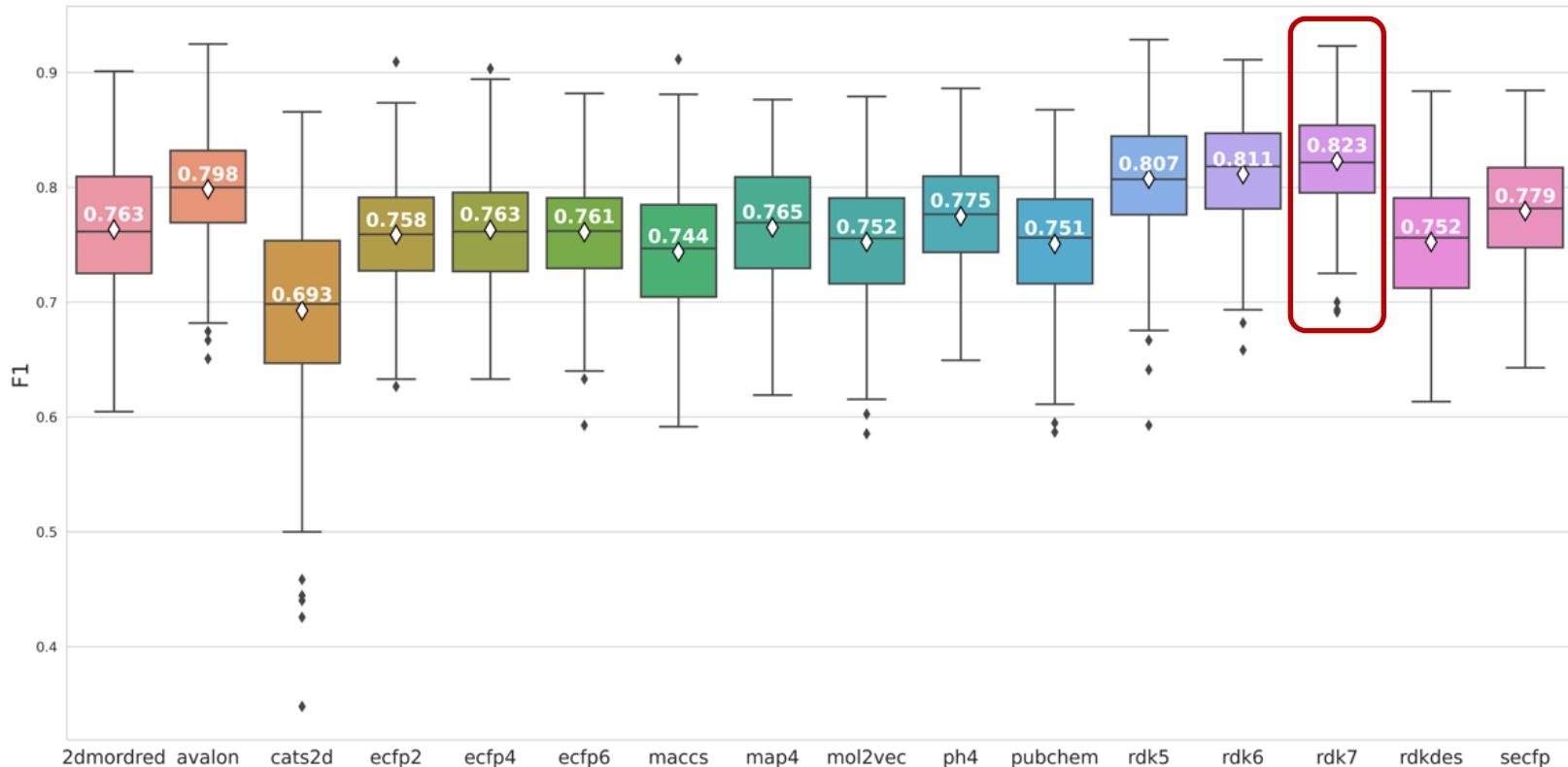
### Data preparation

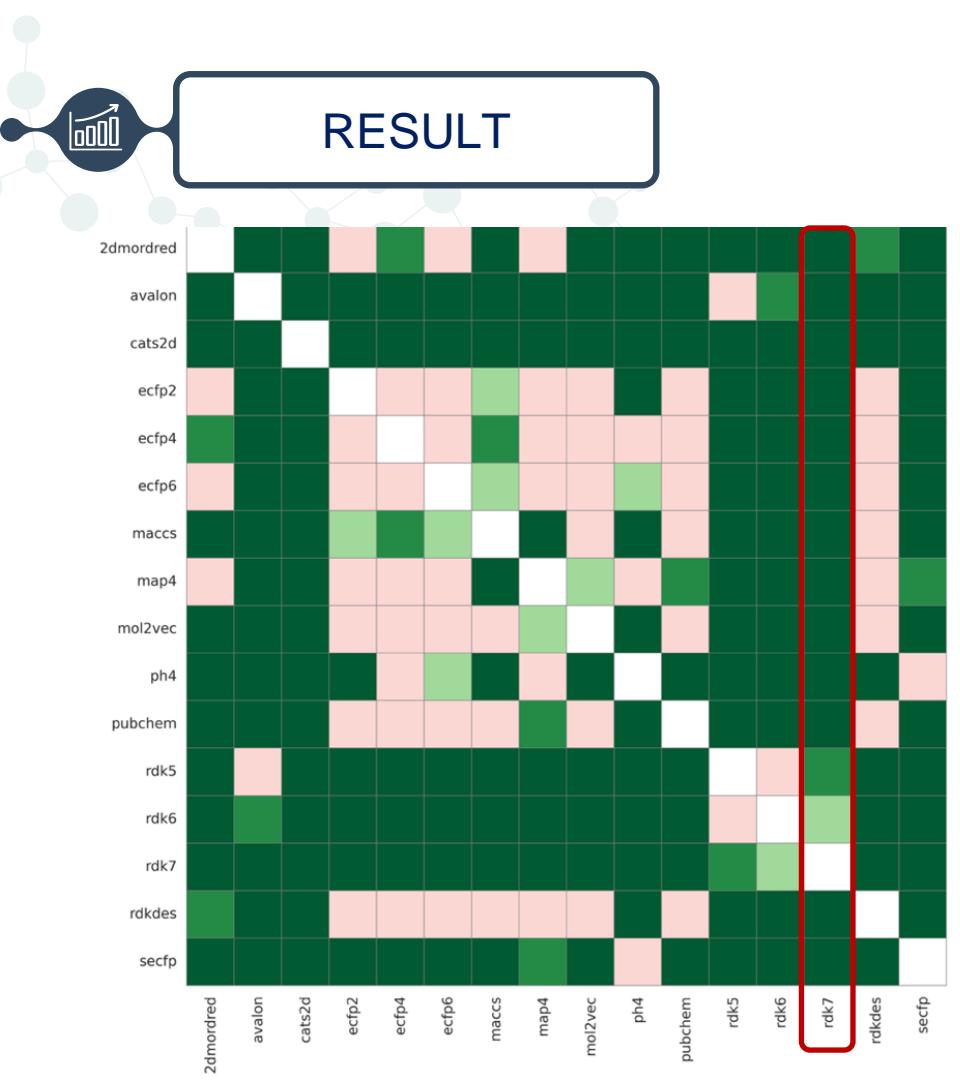


# RESULT

# QSAR – Classification

## Feature selection

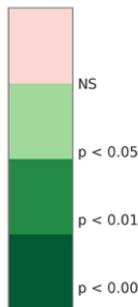




## RESULT

## QSAR – Classification

## Data selection



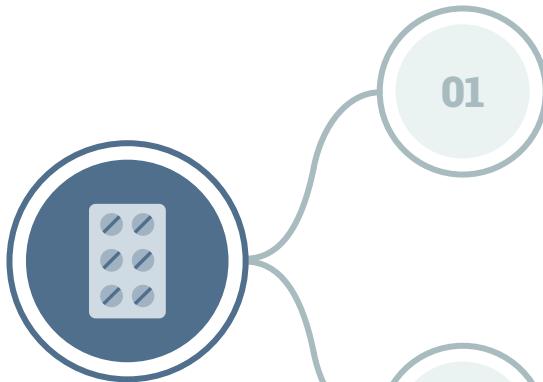
# Primary endpoint

## Highest F1 score



## RESULT

Endpoint



01

Higher F1 score

02

Least features

QSAR – Classification

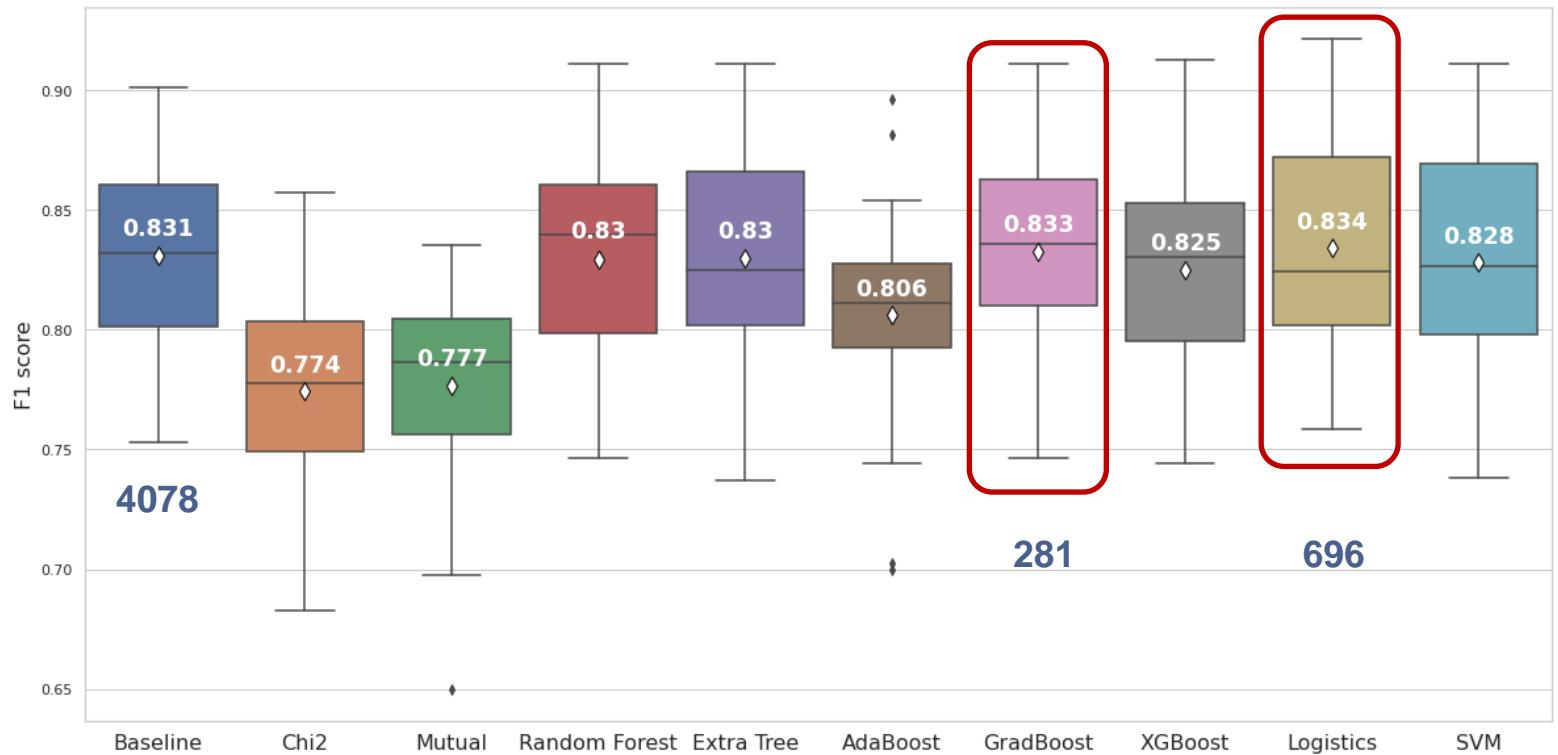
Feature selection



# RESULT

## QSAR – Classification

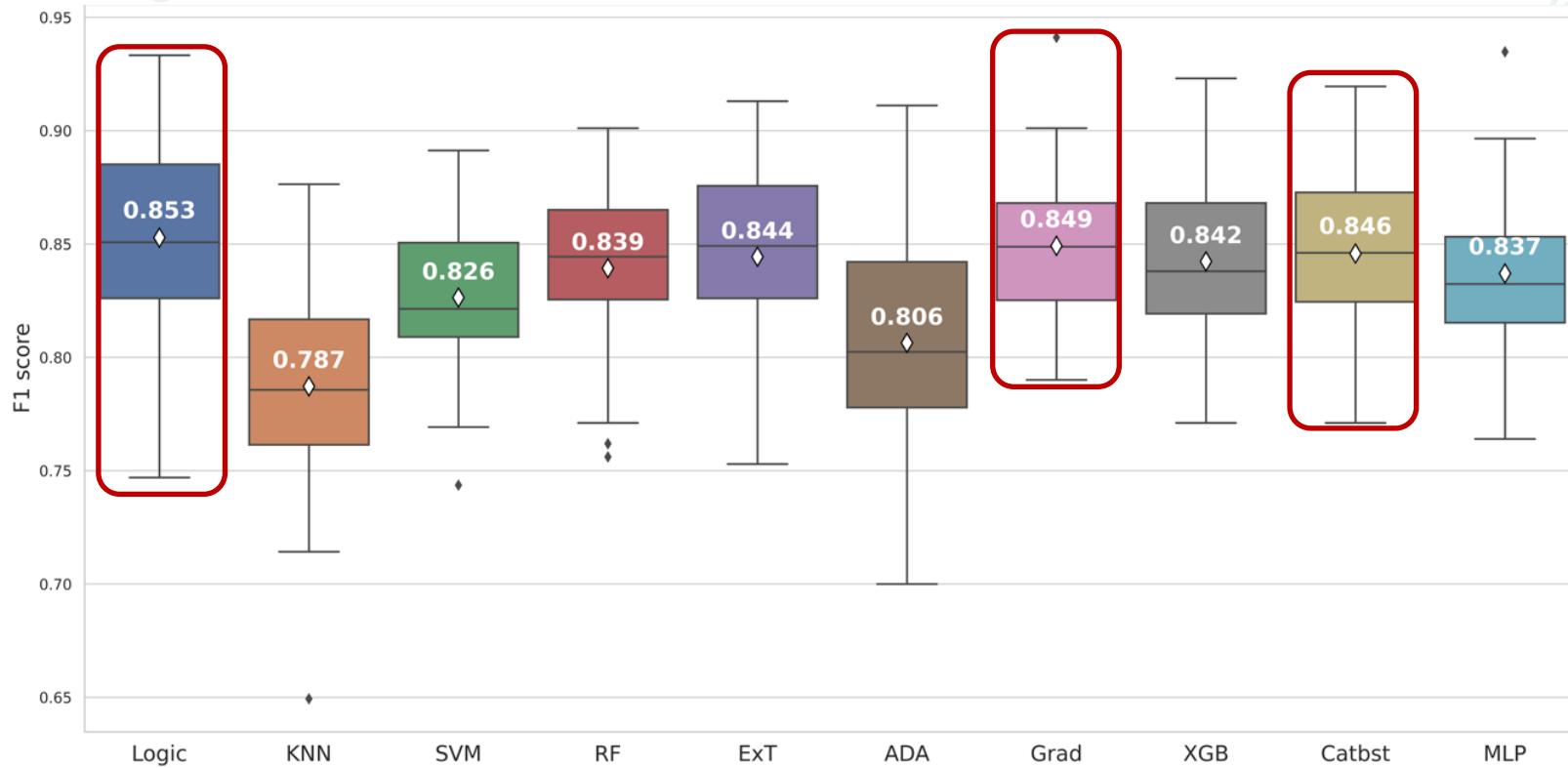
### Feature selection



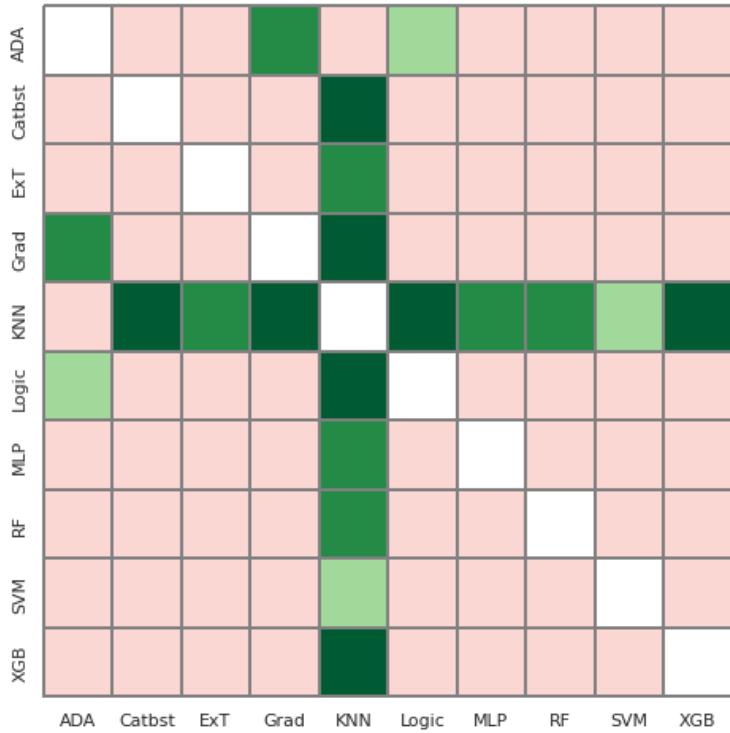
# RESULT

# QSAR – Classification

## Model selection



# RESULT



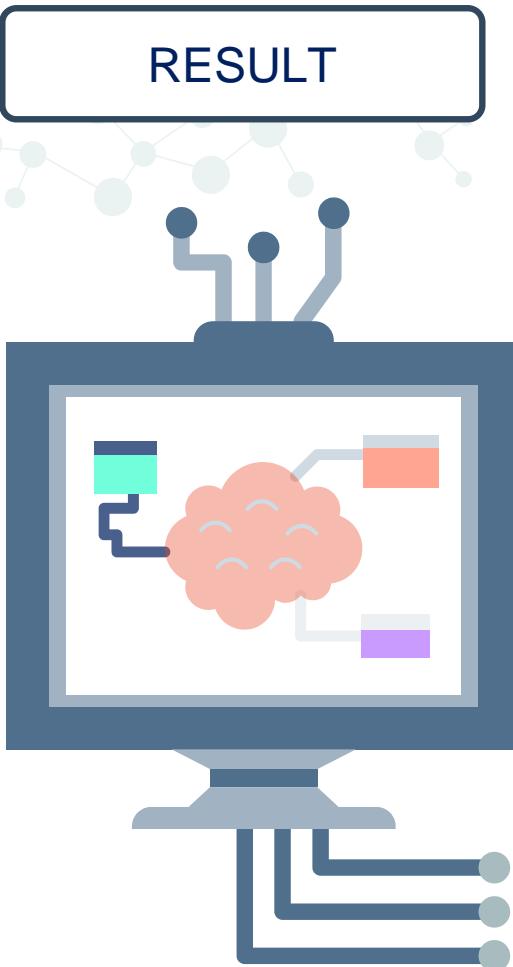
## QSAR – Classification

### Model selection

0

#### Optimized Model

- Logistic regression
- Gradient Boosting
- CatBoost



## QSAR – Classification Model Optimization

### Data sampling

Over-sampling, under-sampling



### Hyperparameter Tuning

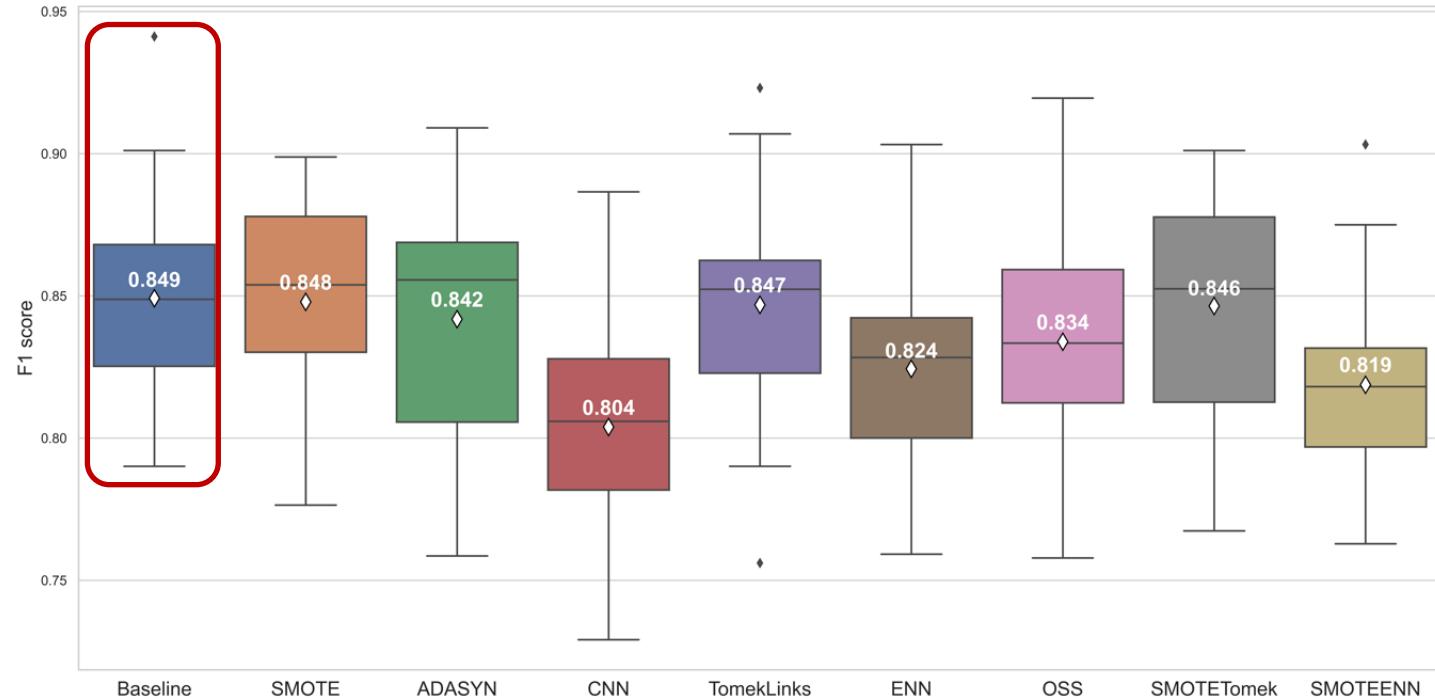
Grid search,

### Probability

Sigmoid (Platt Scaling) or isotonic  
(Isotonic Regression).

# RESULT

## Data sampling

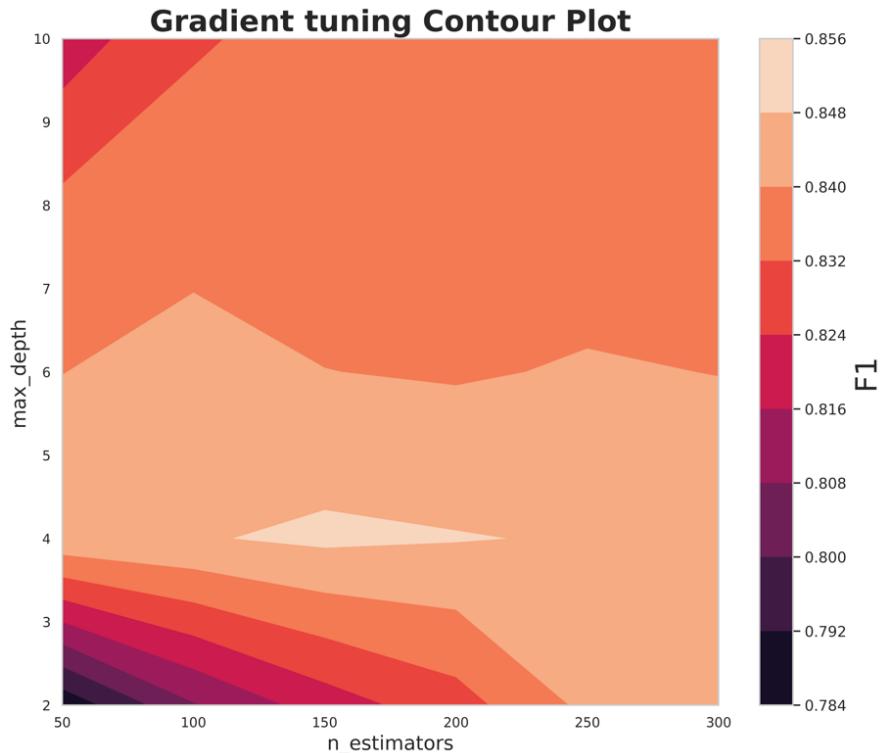


# QSAR – Classification

## Model Optimization

# RESULT

## Tuning

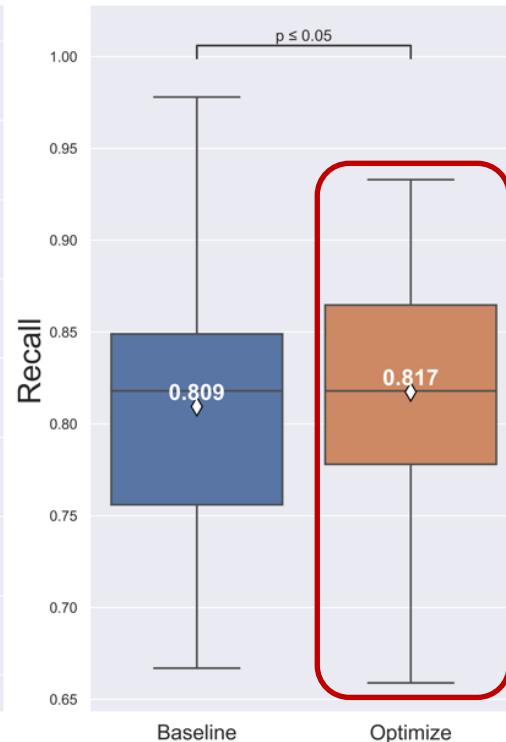
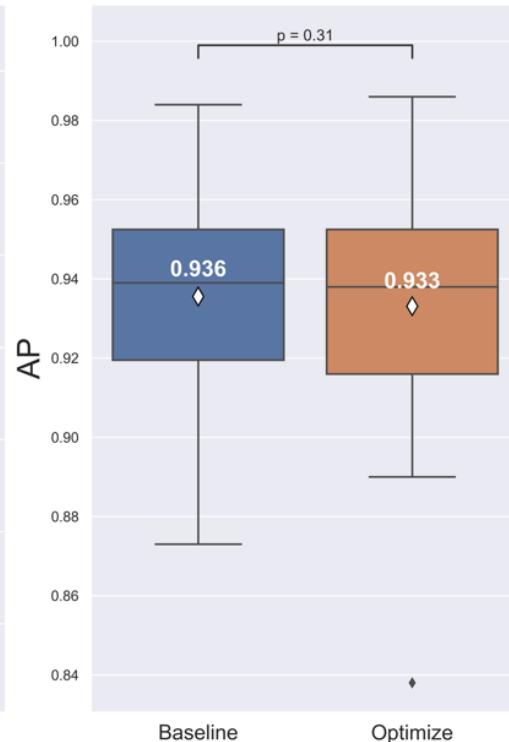
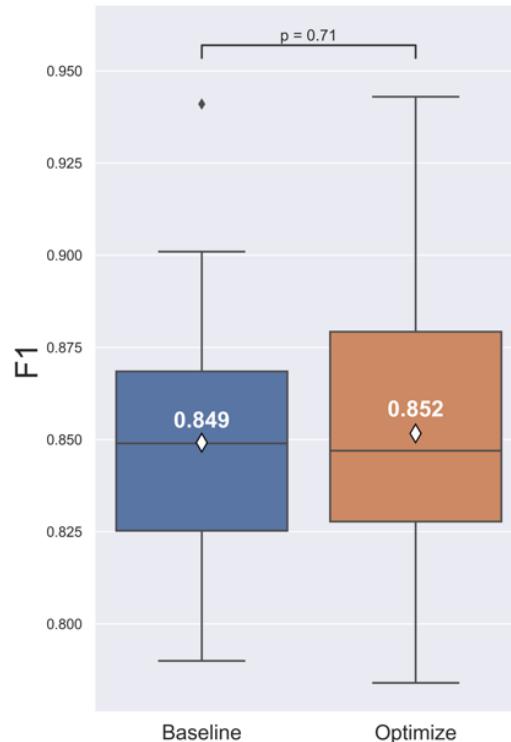


# QSAR – Classification

## Model Optimization



# RESULT



## QSAR – Classification Model Optimization



## RESULT

QSAR – Classification

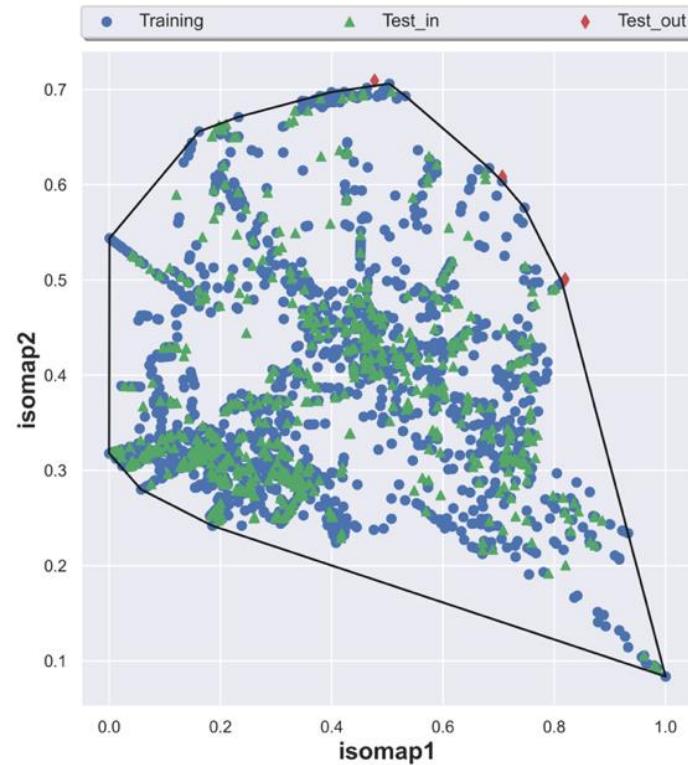
External Validation



Internal Validation			External Validation		
	AP	F1	Độ nhạy	AP	F1
Baseline	0,936	0,849	0,809	0,928	0,873
Optimize	0,933	0,852	<b>0,817</b>	<b>0,938</b>	<b>0,874</b>

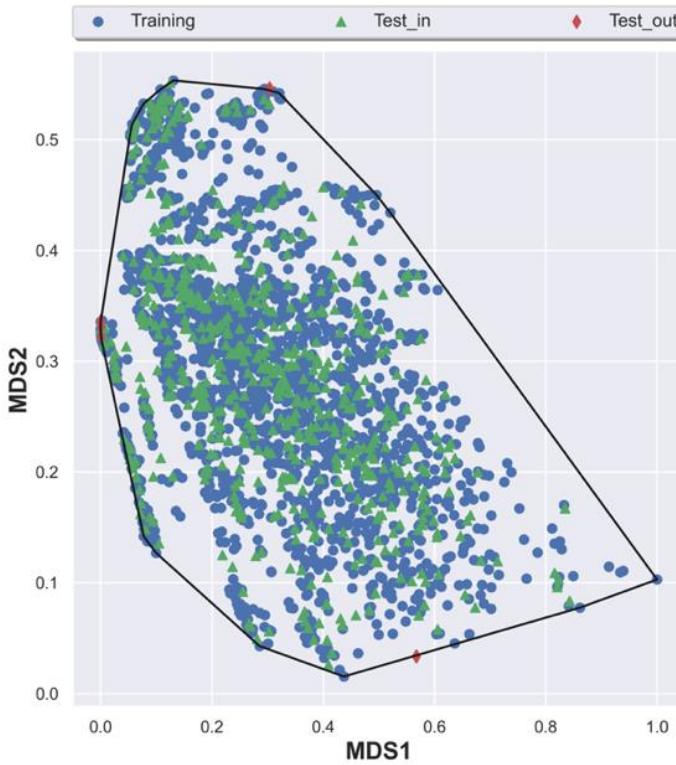
# RESULT

## Convex hull



## QSAR – Classification

## Applicability domain





QSAR – Classification

Applicability domain

## Isomap

CHEMBL394650  
CHEMBL252831  
CHEMBL19043

## MDS

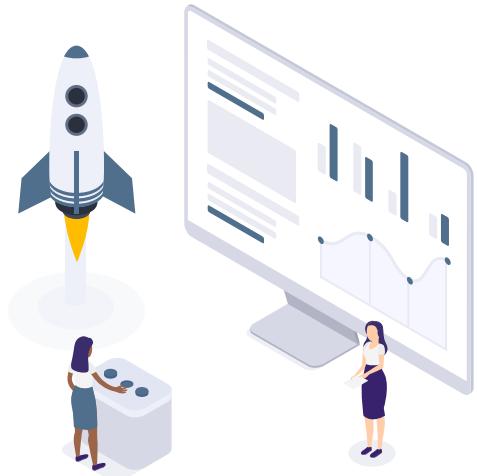
CHEMBL3310415  
CHEMBL2236598  
CHEMBL3310409  
CHEMBL1914564  
CHEMBL252757



RESULT

3

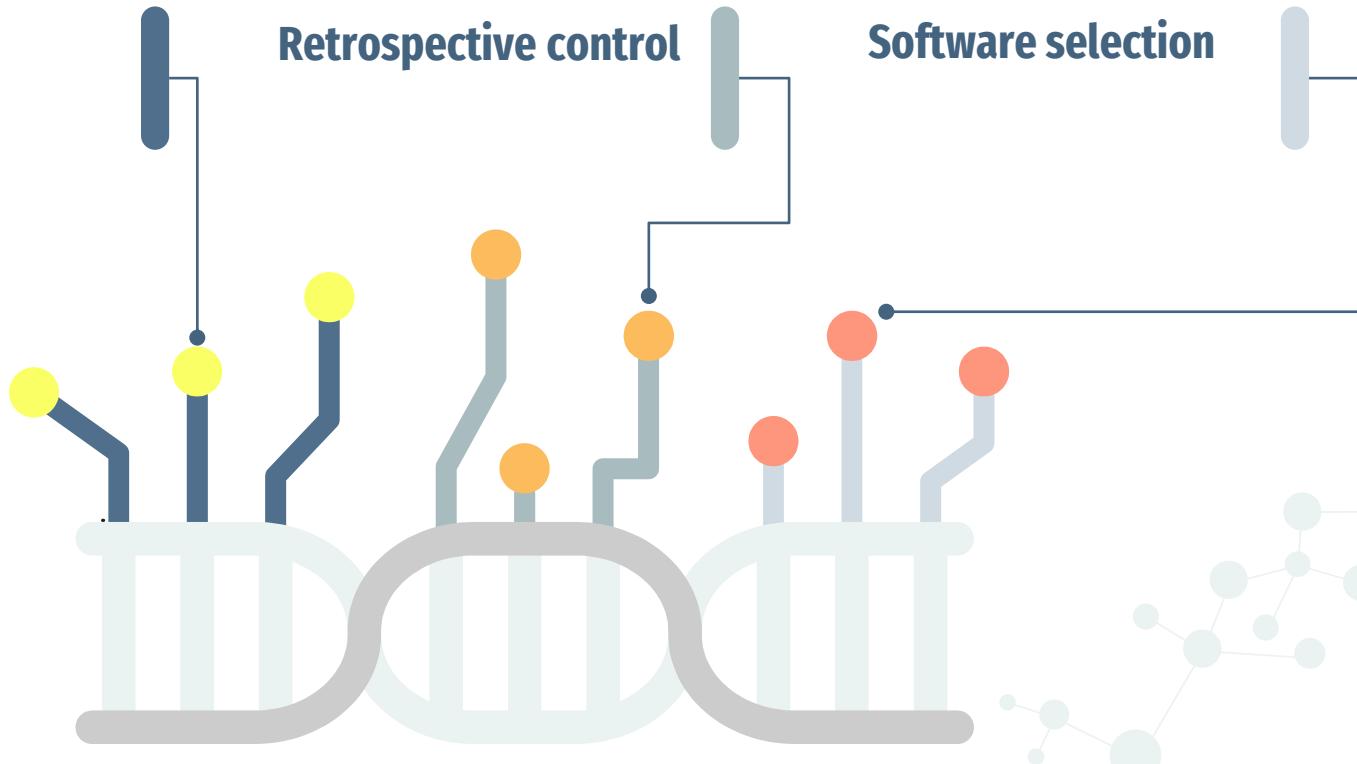
# Docking





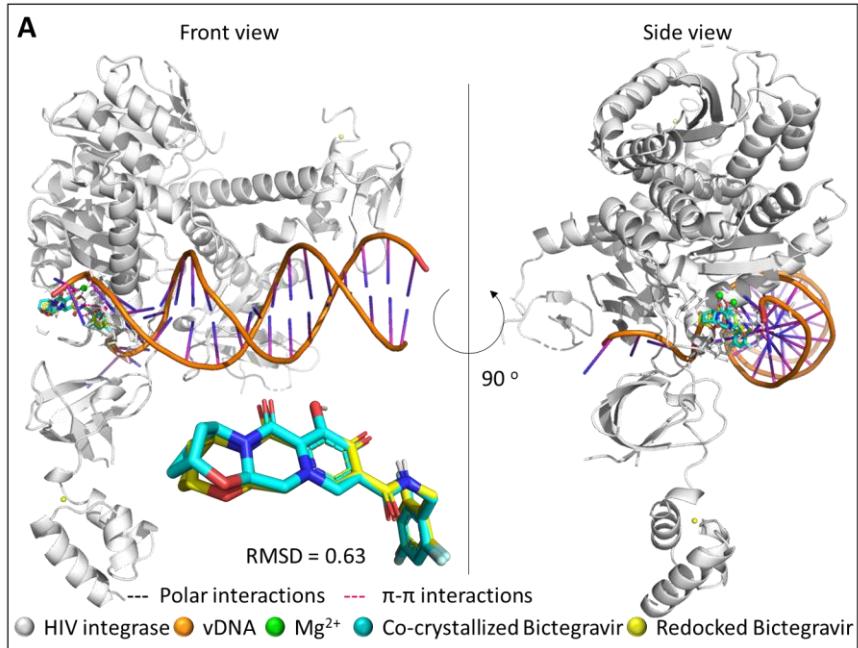
## MOLECULAR DOCKING

**Re-docking**



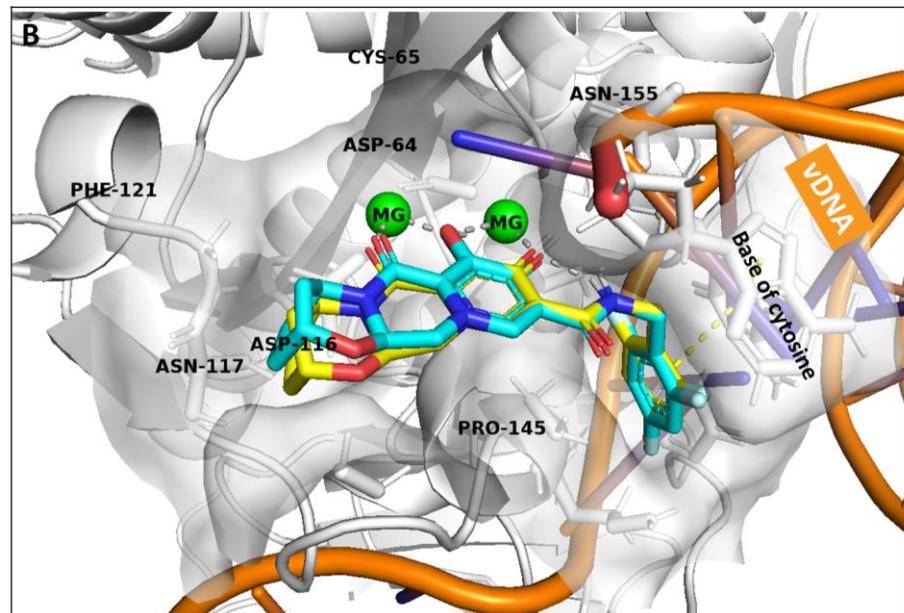
# RESULT

Autodock-GPU



# MOLECULAR DOCKING

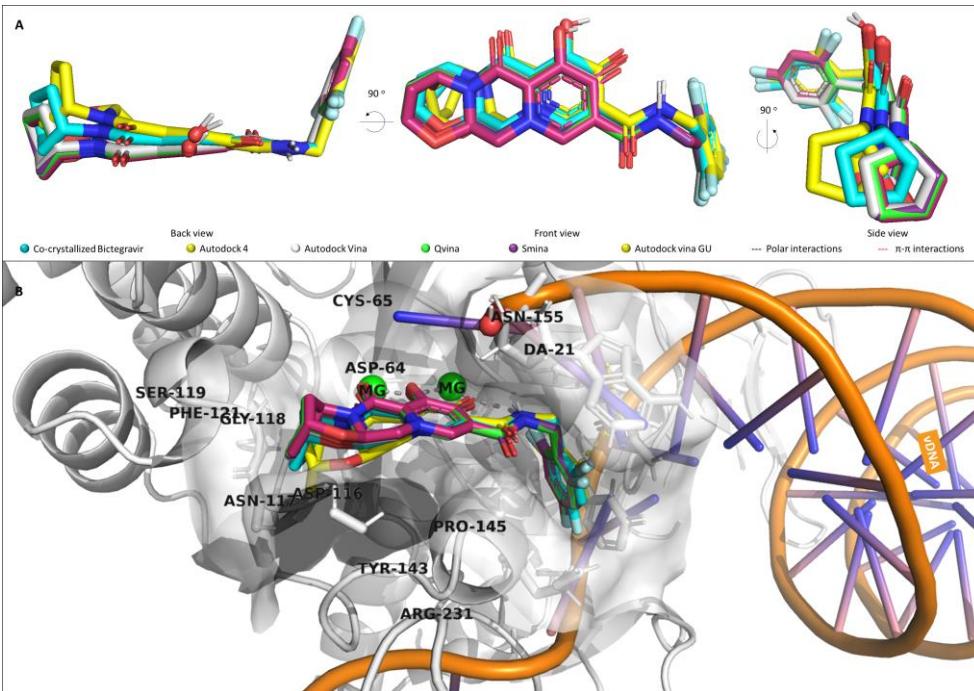
Re-docking



# RESULT

## MOLECULAR DOCKING

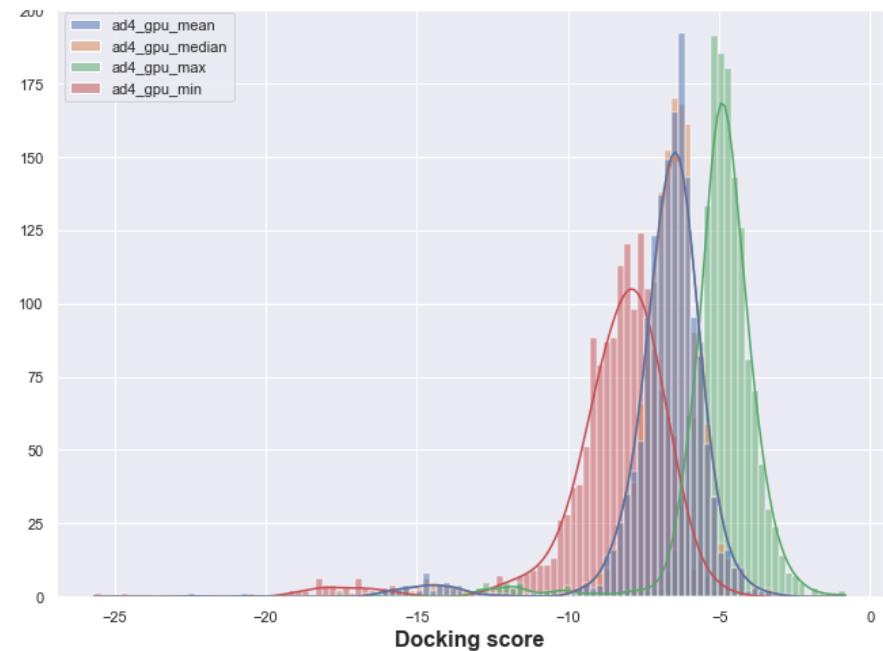
### Re-docking



Softwares	RMSD (Å)	Best docking score(kcal/mol)
<b>Smina</b>	0,677	-10,10
<b>Qvina2</b>	0,716	-9,30
<b>Autodock Vina 1.2.3</b>	0,636	-9,86
<b>Vina-GPU</b>	0,792	-9,80
<b>Autodock-GPU</b>	0,630	-11,11

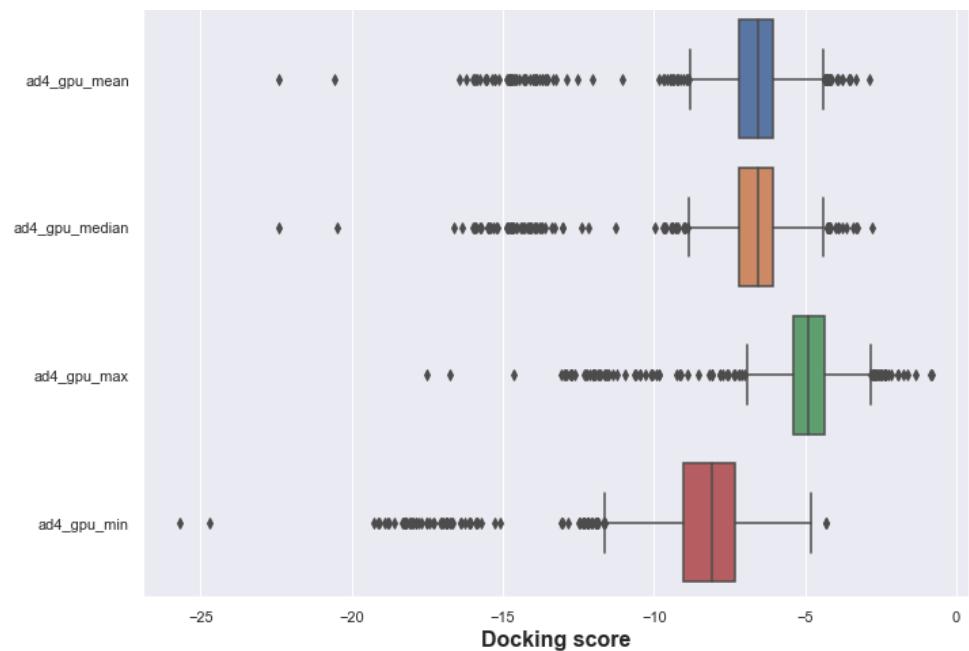
# RESULT

Autodock-GPU



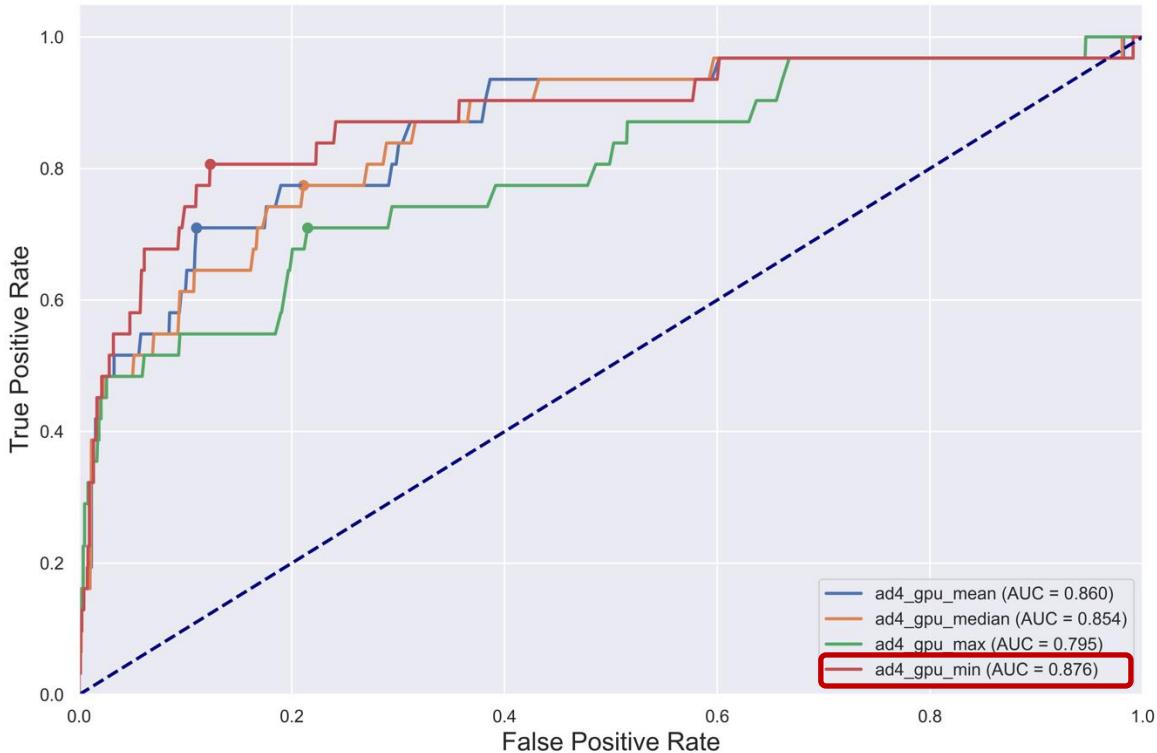
# MOLECULAR DOCKING

Retrospective control



# RESULT

Autodock-GPU



# MOLECULAR DOCKING

Retrospective control

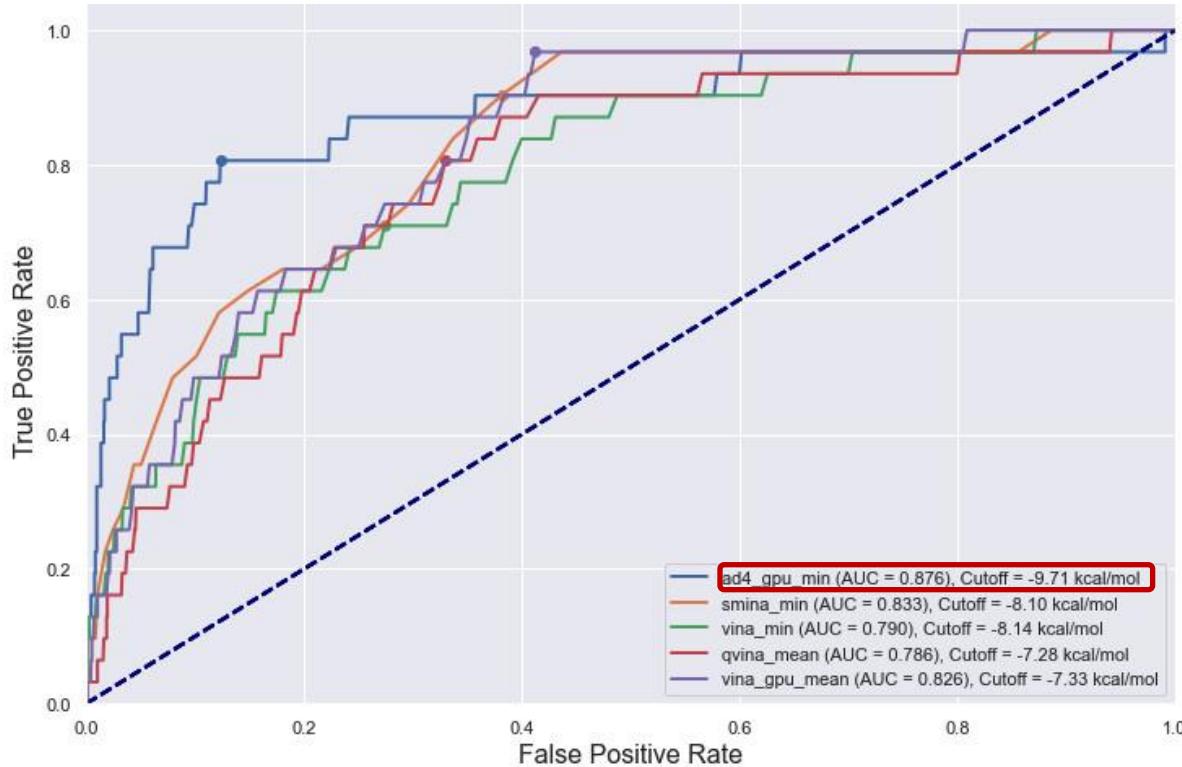


**Autodock-GPU\_min:**

- AUC = 0,876
- G-mean max = 0,841
- TPR = 0,806
- FPR = 0,123
- cutoff = -9,71 kcal/mol

# RESULT

Autodock-GPU



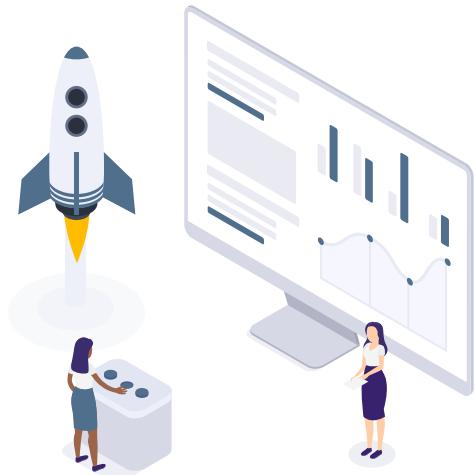
# MOLECULAR DOCKING

Retrospective control

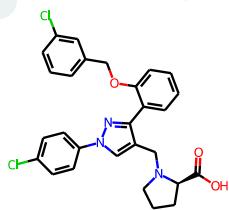


4

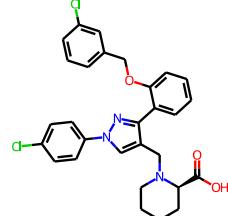
# Screening



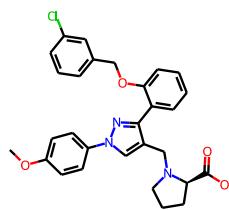
# RESULT



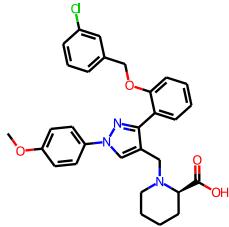
4.3.Pro  
pChEMBL = 7,00



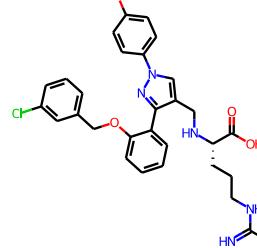
4.3.Pipe  
pChEMBL = 6,92



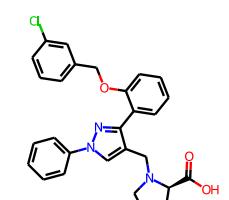
6.3.Pro  
pChEMBL = 7,13



6.3.Pipe  
pChEMBL = 7,08

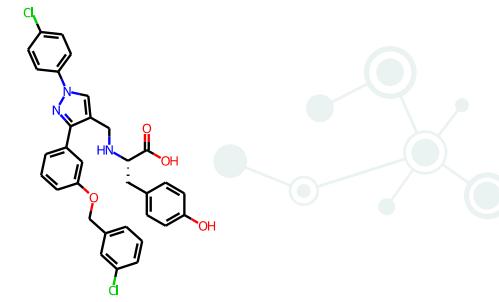


6.3.Arg  
pChEMBL = 7,40



1.3.Pro  
pChEMBL = 7,16

# VIRTUAL SCREENING

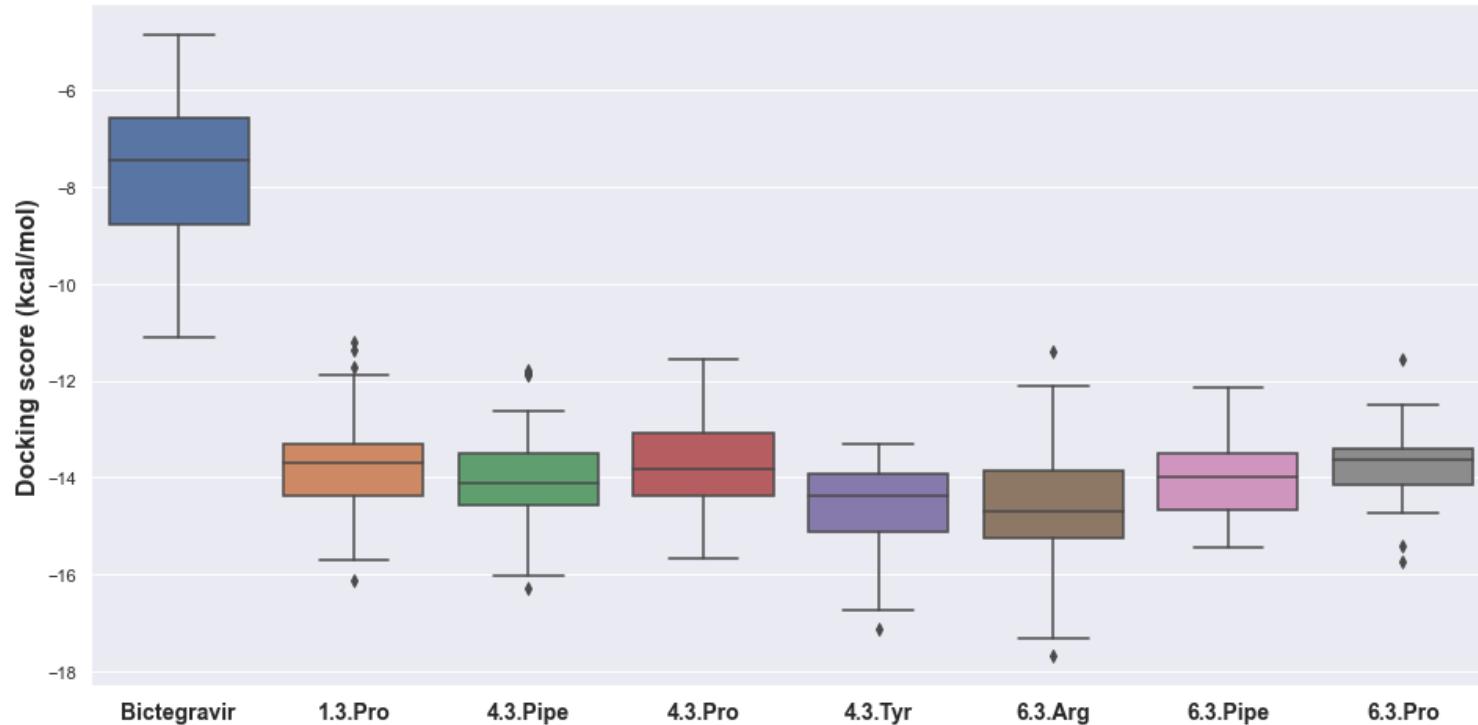


4.3.Tyr  
pChEMBL = 7,06

# RESULT

# VIRTUAL SCREENING

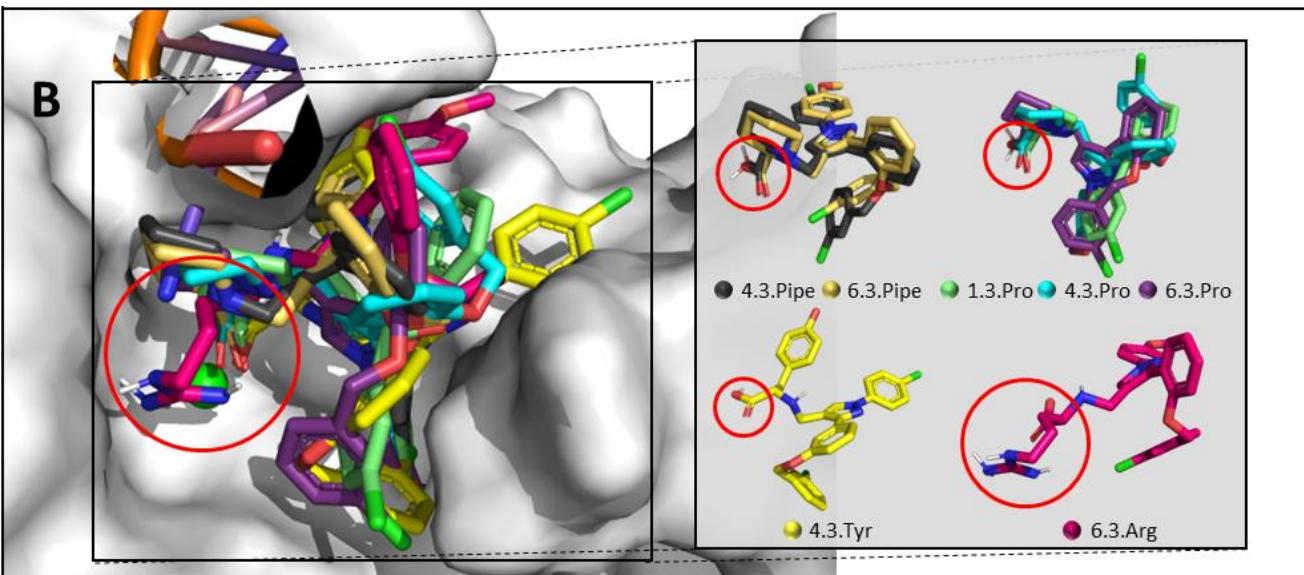
## Molecular Docking



# RESULT

# VIRTUAL SCREENING

## Molecular Docking



**Group 1:** 4.3.Pipe, 6.3.Pipe

**Group 2:** 1.3.Pro, 4.3.Pro,  
6.3.Pro

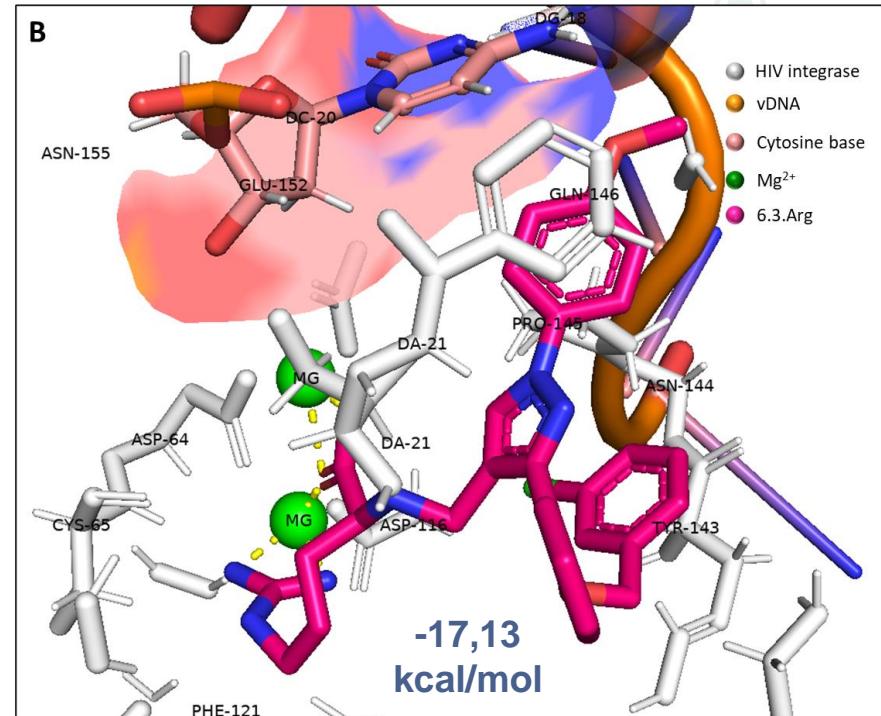
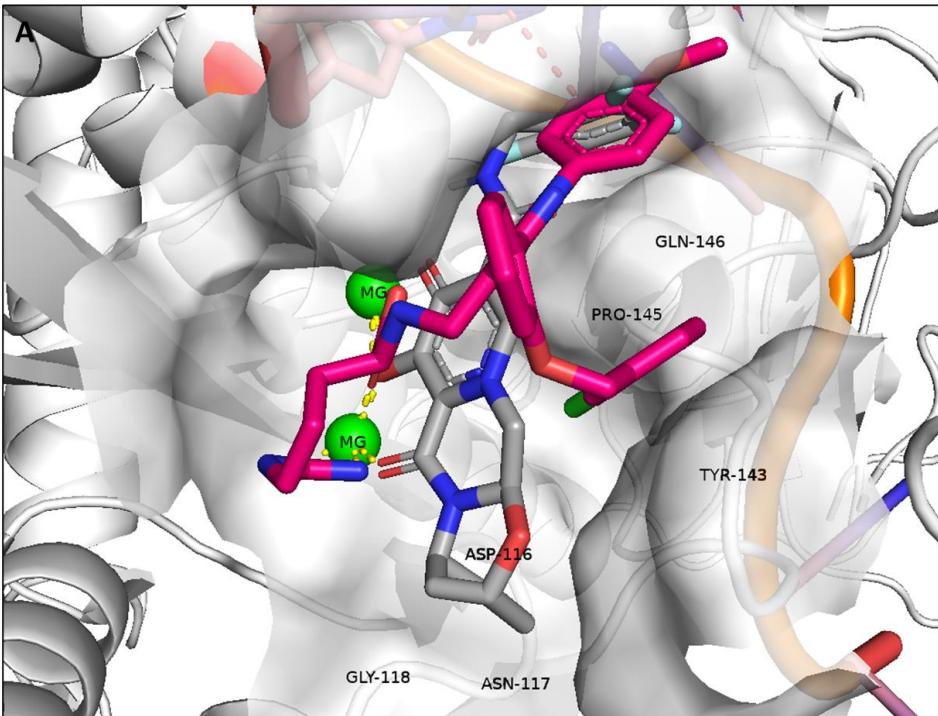
**Group 3:** 4.3.Tyr

**Group 4:** 6.3.Arg



## DISCUSSION

Group 4



## VIRTUAL SCREENING

### Molecular Docking