Accelerating Drug Discovery using Free Energy Perturbation Identification of Potential Inhibitors of SARS-Cov-2 Mpro Cystein Protease

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Introduction

SARS-CoV-2 coronavirus relies on various enzymes to replicate effectively. Two cysteine proteases, namely 3CLpro/Mpro and Plpro are responsible for cleaving the polyproteins, to create the necessary viral proteins for building new viral particles. These proteins have been identified as potential targets for discovering therapeutic agents.

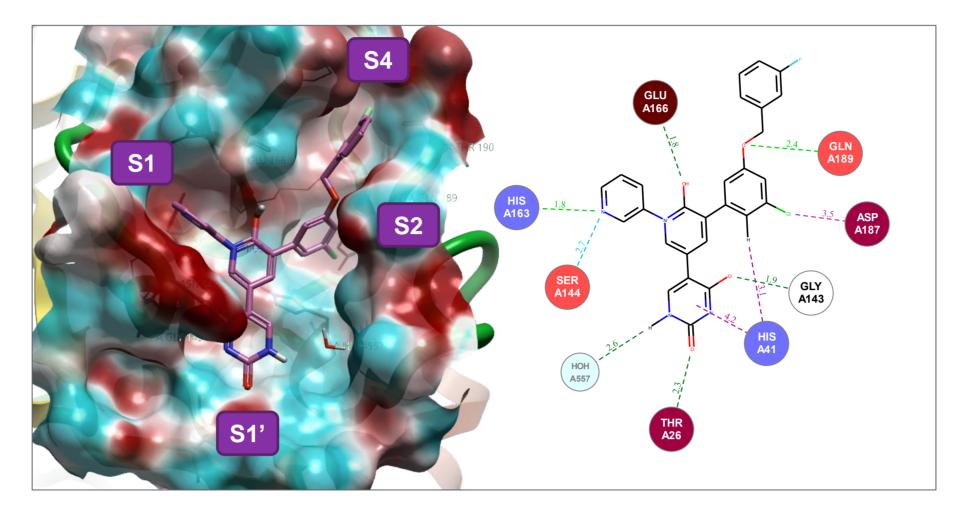
Cresset Discovery, experts in Free Energy Perturbation (FEP) methods, investigated the binding affinity of compounds with known experimental activity to the Mpro protein.¹⁻⁴ The study aimed to elucidate the binding affinity of these compounds to Mpro and to identify the key interactions that contribute to potency.

Protein-Ligand Interactions

The co-crystallized ligand has a cloverleaf motif with a central enol-pyridinone ring:

- Pyridinyl: Pocket S1
- Uracil: Pocket S1'
- Phenyl: Pocket S2

In this study we will explore the contribution of substituents towards the S4 pocket.

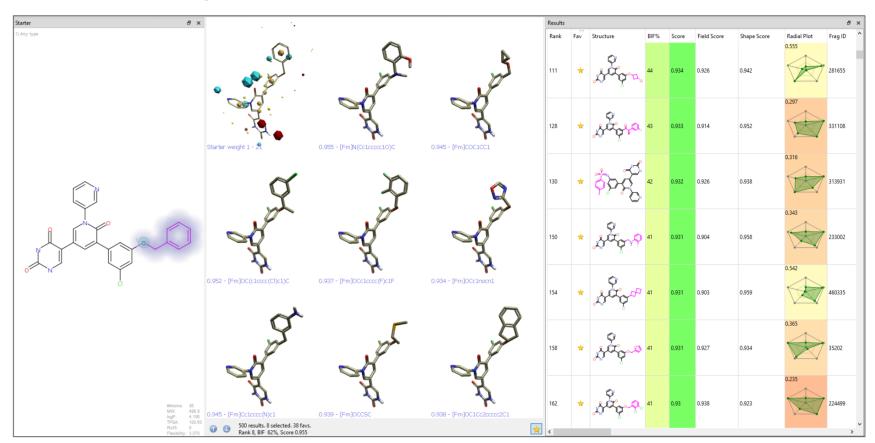


Bioisosteric Replacement

Spark[™], Cresset's bioisosteric replacement software was used to generate new designs for the production run. Spark uses 3D electrostatic and shape properties to explore R-groups on a given scaffold. Results are prioritized based on 3D similarity scores, optionally using a protein as an excluded volume.

Compound 21 was used as a starting point:

- Medium activity ($IC_{50} = 128 \text{ nM}$) •
- Benzylic moiety is of interest for exploration of denovo compounds



Methods

Relative free energies of binding ($\Delta\Delta G$) were obtained with Flare[™] FEP, Cresset's fast and reliable tool for binding affinity predictions, by mutating the ligand in its intermediate states for both the protein-ligand complex in water and the unbound ligand.

- Small molecule forcefield: OpenFF 2.0.0
- Protein forcefield: AMBER FF14SB
- Charge method: AM1-BCC
- Solvent: Explicit TIP3P water
- Initial simulation length per λ window: 4ns

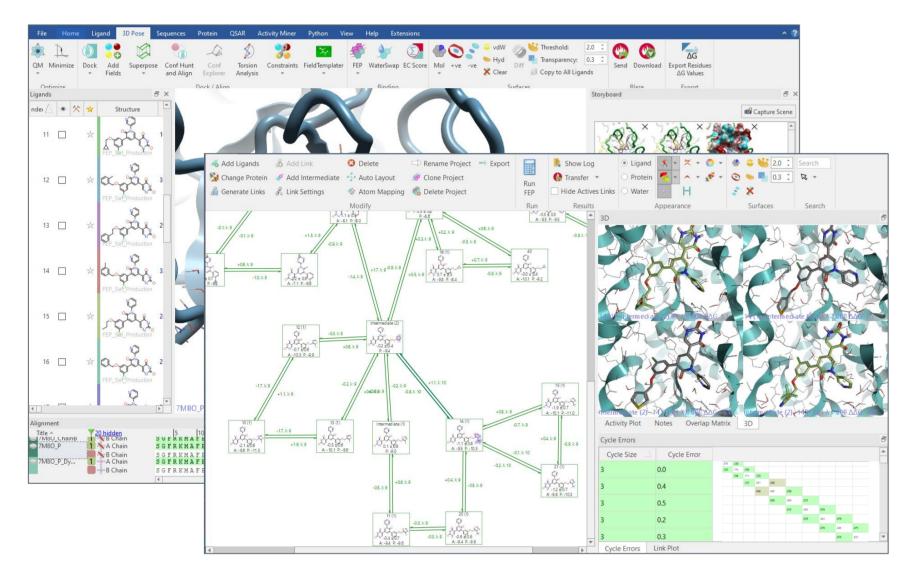


Figure 2: Left: The co-crystallized ligand explores four smaller pockets. Right: Protein- ligand interactions. Purple: Aromatic edgeto-face, Light Purple: Halogen bond, Light/ Dark Green: Strong Hbond/ H-Bond, Light Blue: Weak H-Bond

Benchmark FEP

An evaluation of a total of 54 compounds, with IC50 values reported, was performed to select the most suitable dataset to perform FEP calculations in Flare.

A total of 36 analogues were found more suitable to generate a dataset of compounds that could be used for the FEP calculation.

- Uracil derivatives (28)
- 5-member ring (8)

The 36 compounds were aligned to A YSM 401 (PDB) 7M8O co-crystallized ligand) using the 'Conformation' Hunt and Alignment' method in Flare.

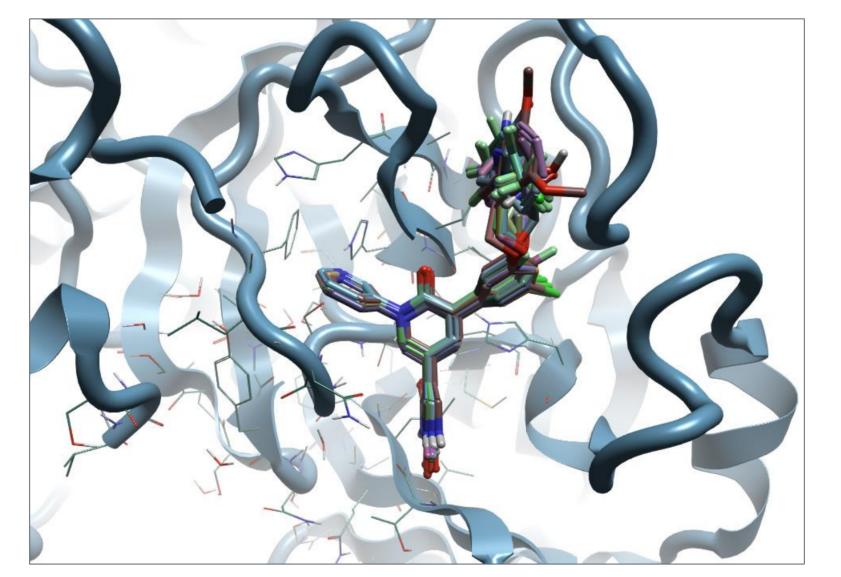
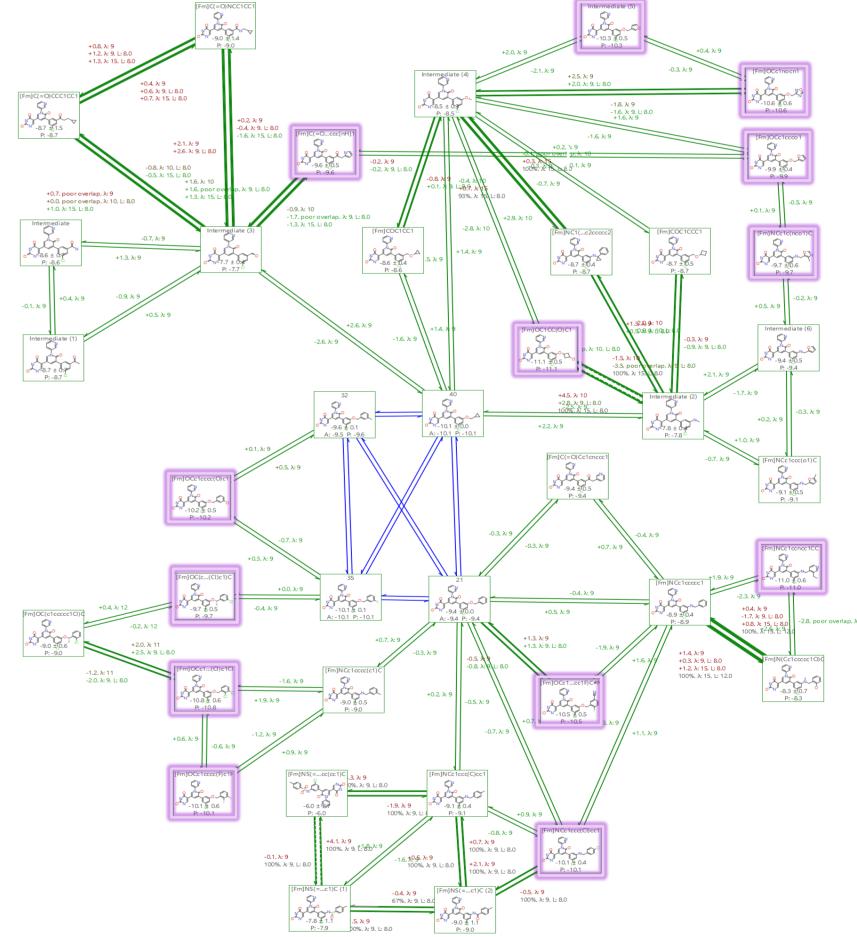


Figure 5: Results of a Spark bioisosteric R-group replacement experiment for compound 21.

From a total of 500 compounds generated, 27 were selected for the FEP prediction run based on: formal charge=0, low 2D similarity towards Compound 21, balanced calculated physicochemical properties, and meaningful chemical modifications for an FEP study.

Production FEP

An FEP Production run with the 27 new compounds and four compounds with known activity was conducted. From the FEP production run, 13 compounds were predicted to be **more active** (10-fold more potent) than the reference Compound 21 ($\Delta G < -9.4$ kcal/mol) and to have similar activity to the most active analogue of the entire dataset (Compound 37, $\Delta G = -10.6$ kcal/mol).



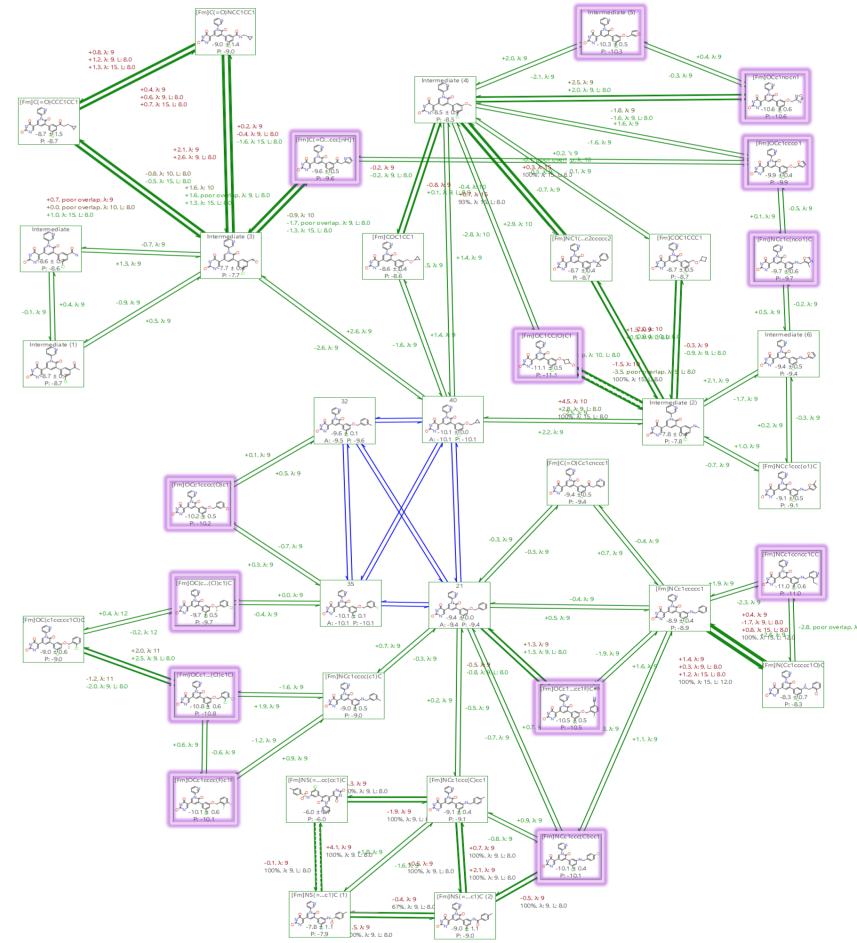


Figure 1: The user-friendly interface of Flare makes FEP calculations easy to run and troubleshoot.

Datasets

A selection of protein crystal structures and cocrystalized ligands were evaluated. The selection was made based on the structure similarity of the ligands.

PDB Code	Resolution	Ligand	IC ₅₀ of Ligand (μΜ)	Type of Protein	Missing loops
7M91	1.95Å	25 A YU4 401	0.025	Monomer: Chain A	Yes
7M90	2.19Å	50	0.25	Monomer: Chain A	Yes
7M8Z	1.79Å	29 A YTV 401	0.25	Monomer: Chain A	Yes
7M8P	2.23Å	23	0.02	Dimer	Yes
7M8O	2.44Å	19 A YSM 401	0.037	Dimer	No

Figure 3: The analogues must be well-aligned in the binding pocket. This can be easily done with the Maximum Common Substructure (MCS) alignment method in Flare.

A well-connected perturbation network with three intermediate molecules was generated for the benchmark FEP run. The FEP graph is generated with LOMAP.⁵

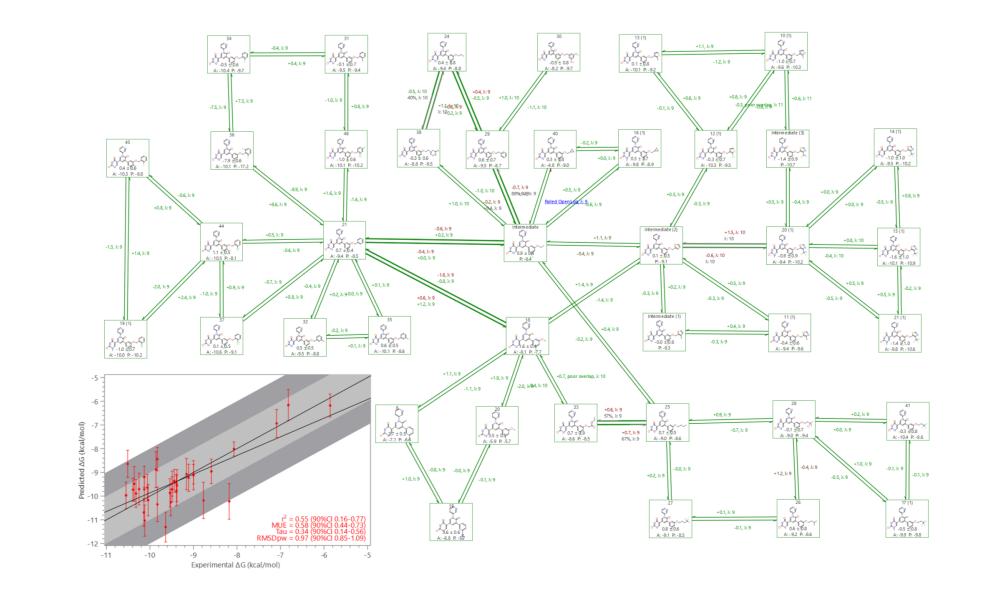


Figure 6: FEP production of 27 compounds generated from Spark, 4 compounds from the benchmark (connected with blue links) and 4 intermediate molecules.

Interesting observations:

• The amino linker is well tolerated compared to the O



Figure 4: FEP perturbation map for the benchmark dataset. On the bottom left, the activity plot shows good correlation between experimental and predicted affinities.

Table 1: Information about crystallographic data associated with this study. The purple box indicates the starting point for our calculations.

PDB 7M8O with co-crystallized ligand A YSM 401 were chosen:

- No missing loops
- Stable Dynamics as monomer and as dimer

Ideally the calculated $\Delta\Delta G$ between ligands should be similar to the difference in the experimental ΔG values, and the calculated ΔG values should be within 1 kcal/mol from the experimentally determined ΔG . In our case good correlation with experimental data and errors below the scientific consensus of 1 kcal/mol were found:

• $R^2 = 0.55$ • MUE = 0.58 kcal/mol • Tau = 0.34 • RMSD = 0.97

S4 pocket prefers halogen substituted phenyl groups

Conclusions

Cresset Discovery's scientists accurately predicted the binding affinity of small molecules to Mpro and generated new designs with predicted high activity. Cresset tools in synergy combined with Cresset Discovery's expertise and knowledge in free energy methods represent a promising approach for accelerating the discovery of new drugs.

References

1. Chun-Hui Zhang, et al., ACS Cent. Sci. 2021, 7, 467-475 2. Chun-Hui Zhang, et al., ACS Med. Chem. Lett. 2021, 12, 1325–1332 3. Maya G. Deshmukh, et al., Structure 29, 823-833, August 5, 2021 4. William L. Jorgensen, WO 2022/150584 A1 5. David L. Mobley, et al., J. Comput. Aided Mol. Des., 2013, 27, 9, 755-770