

#### Check for updates

#### **OPEN ACCESS**

EDITED BY Katarina Polcicova, Slovak Academy of Sciences, Slovakia

\*CORRESPONDENCE Xiaopeng Wei, weixiaopeng@tmu.edu.cn

RECEIVED 29 March 2023 ACCEPTED 22 November 2023 PUBLISHED 15 December 2023

#### CITATION

Gan C, Jia X, Fan S, Wang S, Jing W and Wei X (2023), Virtual screening and molecular dynamics simulation to identify potential SARS-CoV-2 3CL<sup>pro</sup> inhibitors from a natural product compounds library. *Acta Virol.* 67:12464. doi: 10.3389/av.2023.12464

#### COPYRIGHT

© 2023 Gan, Jia, Fan, Wang, Jing and Wei. This is an open-access article distributed under the terms of the Creative Commons Attribution License (CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

# Virtual screening and molecular dynamics simulation to identify potential SARS-CoV-2 3CL<sup>pro</sup> inhibitors from a natural product compounds library

# Chunchun Gan<sup>1</sup>, Xiaopu Jia<sup>2</sup>, Shuai Fan<sup>2</sup>, Shuqing Wang<sup>2</sup>, Weikai Jing<sup>2</sup> and Xiaopeng Wei<sup>2</sup>\*

<sup>1</sup>School of Medicine, Quzhou College of Technology, Quzhou, China, <sup>2</sup>Tianjin Key Laboratory on Technologies Enabling Development of Clinical Therapeutics and Diagnostics, School of Pharmacy, Tianjin Medical University, Tianjin, China

Based on the crystal structure of the 3C-like protease/Nsp5 (PDB ID 6W63), virtual hits were screened from a natural product compounds database-containing 407270 natural products-by using the highthroughput virtual screening (HTVS) module of Discovery Studio software, and then filtering by "Lipinski's rule of five" from the top 20 virtual hits. Two star-hits were selected by CDOCKER results and the protein-ligand interactions with the 3CL<sup>pro</sup> were analyzed. Finally, a 100 ns molecular dynamics simulation was carried out to verify the stability of the receptor-ligand complexes. We screened potent broad-spectrum non-covalent inhibitors that could bind to the SARS-CoV-2 3CL<sup>pro</sup> active binding site from the natural product compounds library through HTVS and molecular dynamics simulations methods. The LibDock scores and -CDOCKER energy value of the star-hits were higher than the original ligands (X77) bound to 3CL<sup>pro</sup>. CNP0348829 and CNP0474002, as star-hits, can bind stably to the active site of 3CL<sup>pro</sup>, which are promising candidate compounds for the treatment of SARS-CoV-2 and provide a theoretical basis for the development of antiviral drugs. The results of the present study may be useful in the prevention and therapeutic perspectives of COVID-19. However, further in vitro and in vivo validation tests are required in the future.

#### KEYWORDS

SARS-CoV-2,  $3CL^{pro}$ , natural product compounds, virtual screening, molecular docking

# Introduction

In December 2019, an outbreak of a severe respiratory disease occurred in Wuhan, Hubei Province, China. On 11 February 2020, the World Health Organization announced that the new coronavirus pneumonia was named "COVID-19" (World Health Organization, 2022). The International Committee on Classification of Viruses Gan et al.



has named the new coronavirus "SARS-CoV-2" (Wang et al., 2020). It is a zoonotic pathogen belonging to the genus  $\beta$ -coronavirus ( $\beta$ -CoV) of the Coronaviridae subfamily of ortho-coronaviruses. The virus initially attacks the respiratory system and causes flu-like symptoms such as cough and fever, and in severe cases, patients may develop acute respiratory distress syndrome and respiratory failure (Sahebnasagh et al., 2020). In addition, novel coronavirus pneumonia can cause systemic inflammation and acute cardiac damage, leading to arrhythmias, heart failure, and multi-organ dysfunction in critically ill patients (Wang et al., 2020). As of December 2022, over 645 million confirmed cases, and over 6.6 million deaths, have been reported globally.<sup>1</sup>

Nsp5, sometimes referred to as 3C-like protease (3CL<sup>pro</sup>) or the major protease, is a cysteine protease that breaks down polyproteins at 11 different locations to produce mature, functional proteins. Viral replication depends on 3CL<sup>pro</sup>, the inhibition of which limits the synthesis of key enzymes for replication like RNA-dependent RNA polymerase (Ullrich and Nitsche, 2020). Furthermore, as the 3CL<sup>pro</sup> and spike proteins are independent proteins encoded in different parts of the viral genome, the 3CL<sup>pro</sup> inhibitor's antiviral activity would probably not be impacted and it does not trigger mutations in the spike protein, which frequently occur in SARS-CoV-2 variations. Consequently, 3CL<sup>pro</sup> is a potential target for COVID-19 treatment using small-molecule oral therapies (Huang et al., 2021; Unoh et al., 2022).

Natural products have always played a crucial role in drug development against various diseases. Therefore, traditional herbs from diverse geographical locations and various habitats could be considered as potential sources of new drugs for the treatment of viral infections, including those caused by SARS-CoVs and its emergent mutants. After the outbreak of novel coronavirus pneumonia, the National Health Commission of People's Republic of China promptly issued the "Treatment Protocol for COVID-19," which has been updated to the ninth edition on a trial basis as of 15 March 2022. In the course of COVID-19 treatment, Traditional Chinese Medicine (TCM) including "XueBiJing injection," "TanReQing injection," "JinHuaQingGan granule," "XiYanPing injection" and others participated in the whole treatment process and played a significant role (Zhuang et al., 2020). On 31 March 2022, the World Health Organization<sup>2</sup> published a report stating that

<sup>1</sup> https://www.who.int/publications/m/item/weekly-epidemiologicalupdate-on-covid-19–14-december-2022

<sup>2</sup> https://www.who.int/publications/m/item/who-expert-meeting-onevaluation-of-traditional-chinese-medicine-in-the-treatment-ofcovid-19

Coconut ID	Structure	LibDOCK scores	-CDOCKER_ENERGY (kcal/mol)
CNP0465303		188.883	-32.3854
CNP0455361	$ \begin{array}{c} \begin{array}{c} \\ \\ \\ \end{array} \end{array} \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ $	183.594	-12.7831
CNP0457047		178.976	14.2185
CNP0139709		167.513	26.1017
CNP0458104		166.682	-13.015
CNP0478503	N N N N N N N N N N N N N N N N N N N	159.788	-7.49867
CNP0475158	HO, N N N O O OH O OH O	156.399	9.84003
CNP0447420		155.407	23.1654
CNP0457847	HO HO HO HO HO HO HO HO HO HO HO HO HO H	154.03	-13.7951

## TABLE 1 The LibDOCK scores and CDOCKER\_ENERGY of top 20 virtual hits.

(Continued on following page)

Coconut ID	Structure	LibDOCK scores	-CDOCKER_ENERGY (kcal/mol)
CNP0478727	OH O O'''OH	150.303	-34.5301
CNP0446191	HN, HN, HH2 HN, HN, HH2	150.107	24.5206
CNP0474861		143.167	-52.9031
★CNP0348829	O C C C C C C C C C C C C C C C C C C C	142.061	38.6734
CNP0475639		140.811	-6.46299
CNP0475126		138.716	19.2972
★CNP0474002		138.328	35.6722
CNP0474130		138.274	4.57313
CNP0458286	$H_2N - \begin{pmatrix} 0 \\ 0 \\ H_N \\ H_N \\ N \\ \end{pmatrix} = N$	137.822	6.98237

TABLE 1 (Continued) The LibDOCK scores and CDOCKER\_ENERGY of top 20 virtual hits.

(Continued on following page)

Coconut ID	Structure	LibDOCK scores	-CDOCKER_ENERGY (kcal/mol)
CNP0474134		137.745	13.4454
CNP0436312		136.109	-19.9438
Inhibitor (X77)		123.888	30.4747

TABLE 1 (Continued) The LibDOCK scores and CDOCKER\_ENERGY of top 20 virtual hits.

\*Screened star-hit compounds are highlighted with bold font and with the star symbol.

"TCM can effectively treat COVID-19, reduce the conversion of mild and common cases to severe disease, shorten the time of virus clearance, and improve the clinical prognosis of mild and common patients" (WHO Team, 2022), which shows that natural products play a very important role in the prevention and treatment of COVID-19. Therefore, screening for candidates from a natural products library<sup>3</sup> is a potential strategy for antiviral drug development.

# Materials and methods

## High-throughput virtual screening

The structures of compounds were obtained from the Coconut database,<sup>3</sup> and the crystal structure of 3CL<sup>pro</sup> (6W63) was downloaded from the PDB database.<sup>4</sup> Discovery Studio software (version 4.0, Biovea Inc., Omaha, NE, USA) was applied to molecule docking. The compounds of the Coconut database were prepared using the "Prepare Ligands" module. After removing crystallographic water molecules and ligands, the "Prepare Protein" module was used to remove the conformation of the target protein and to fill the incomplete amino acid residues. Then, the virtual screening was performed in the

"LibDock" module, which is a virtual high-throughput algorithm for docking ligands into the receptor active sites. CDOCKER, as an implementation of a CHARMm based docking tool using a rigid receptor, was used to further molecular docking. Ligand conformations were aligned to receptor interaction sites (hotspots) and the best scoring poses were reported (Wu et al., 2003; Mohan et al., 2015).

### Molecular dynamics (MD) simulation

In order to verify the time-dependent stability of ligandreceptor complexes, Desmond\_maestro 2020.4 with OPLS\_ 2005 force filed<sup>5</sup> tools was used to perform a 100-ns molecular dynamics simulation on the receptor-ligand complexes obtained by virtual screening (Okimoto et al., 2009; Gahlawat et al., 2020). Firstly, in an orthorhombic box, the system was solvated with the SPC water model and neutralized by Cl<sup>-</sup> or Na<sup>+</sup> ions. The system energy was minimized at T = 0 K while the force field could be used to regulate the structure of the molecules, followed by ensuring the system stability by the equilibration processes and heating. Subsequently, MD simulation was carried out under periodic boundary conditions and NPT ensemble at T = 300 K and 1 atm pressure using a Nose-Hoover

<sup>3</sup> https://coconut.naturalproducts.net/

<sup>4</sup> https://www.pdbus.org/

<sup>5</sup> https://www.deshawresearch.com/downloads/download\_ desmond.cgi/

Compound <sup>a</sup>	Formula	$MW^{b}$	NRB <sup>c</sup>	<b>NBD</b> <sup>d</sup>	NBA <sup>e</sup>	MLogP <sup>f</sup>	GI absorption <sup>g</sup>
CNP0465303	$C_{49}H_{48}N_2O_7$	776.915	10	4	9	8.082	Low
CNP0455361	$C_{50}H_{68}N_2O_6$	792.510	9	3	7	4.406	Low
CNP0457047	C35H37NO7	583.671	10	4	7	6.033	Low
CNP0139709	$C_{30}H_{42}O_7$	514.65	10	3	7	5.769	High
CNP0458104	$C_{30}H_{37}N_{1}O_{5} \\$	491.618	10	1	6	4.642	High
CNP0478503	$C_{25}H_{30}N_{2}O_{4} \\$	423.525	9	3	5	1.499	High
CNP0475158	$C_{27}H_{45}N_3O_5$	491.663	10	4	6	1.276	Low
CNP0447420	$C_{23}H_{25}N_5O_5$	567.628	6	1	8	3.745	Low
CNP0457847	$C_{33}H_{49}N_1O_6$	555.745	8	5	6	4.94	Low
CNP0478727	$C_{40}H_{42}N_4O_5$	659.793	9	4	8	3.544	Low
CNP0446191	$C_{22}H_{30}N_4O_5$	430.497	7	6	6	-0.859	Low
CNP0474861	$C_{29}H_{36}O_{9}$	528.591	3	2	9	1.393	Low
★CNP0348829	$C_{24}H_{27}N_1O_5$	409.475	8	2	5	3.742	High
CNP0475639	$C_{24}H_{45}N_3O_3$	425.648	6	5	4	-1.863	Low
CNP0475126	$\mathrm{C}_{22}\mathrm{H}_{31}\mathrm{FN}_4\mathrm{O}_3$	418.505	6	3	5	-0.069	High
★CNP0474002	$\mathrm{C}_{24}\mathrm{H}_{33}\mathrm{FN}_4\mathrm{O}_5$	476.541	8	2	5	1.164	High
CNP0474130	$C_{23}H_{19}N_5O_2$	479.515	5	3	4	3.811	High
CNP0458286	$C_{19}H_{19}N_5O_4$	381.385	6	2	8	0.419	High
CNP0474134	$C_{20}H_{28}N_6O_2S$	416.54	9	4	5	0.3	High
CNP0436312	$C_{29}H_{44}O_8$	520.655	10	0	8	3.81	High
Inhibitor (X77)	C <sub>27</sub> H <sub>33</sub> N <sub>5</sub> O <sub>2</sub>	459.583	7	2	4	3.886	High

#### TABLE 2 ADME properties of top 20 virtual hits.

<sup>a</sup>Compound ID from Coconut database.

<sup>b</sup>Molecular weight of the compounds (acceptable range: ≤500 g/mol).

<sup>c</sup>Number of rotatable bonds (acceptable range:  $\leq$ 10).

<sup>d</sup>Number of H-bond donors (acceptable range:  $\leq$ 5).

<sup>e</sup>Number of H-bond acceptors (acceptable range: <=10). <sup>f</sup>MlogP: log P by the method of Moriguchi (acceptable range: 4.5).

whogi, log i by the method of wongden (acceptable rang

<sup>g</sup>Gatrointestinal absorption.

thermostat and Martyna-Tuckerman-Klein barostat (Li et al., 2019; Srivastava et al., 2021). The parameters, such as root-meansquare-deviation (RMSD) for the receptor and ligand, rootmean-square-fluctuation (RMSF) of the protein and ligand, or numbers of hydrogen bonds, were calculated to evaluate the stability of receptor-ligand complexes. The change in conformation can be analyzed by the change in RMSD, showing whether the equilibrium is reached during the simulation. When the shift between the specific frame and the reference frame is 1–3 Å, it can be considered that the shift of the specific frame is not much different from the reference frame. However, it indicates that a larger conformational change has occurred during the simulation when the shift change is apparent. The backbone RMSD curve represents the RMSD of the protein backbone over time. Lig\_fit\_Prot curve represents the RMSD of the ligand when the protein-ligand complex is first aligned to the protein backbone and then the RMSD of the heavy atom is measured. It is likely that the ligand has diffused away from its initial binding site if a value significantly larger than the RMSD of the protein is observed. The lig\_fit\_lig curve represents the RMSD of the ligand when the ligand is aligned in its reference conformation which is used to measure the internal fluctuations of the ligand atoms.

## ADMET properties prediction

AdmetSAR2<sup>6</sup> is a website which allows user to search for ADME/T (Absorption, Distribution, Metabolism, Excretion, and Toxicity including Ames toxicity, carcinogenic properties, acute oral toxicity, and rat acute toxicity) properties profiling by name, CASRN, and similarity search (Yang et al., 2019). At the same time, Swiss-ADME online web server<sup>7</sup> was used to predict physicochemical properties, water solubility, pharmacokinetics, lipophilicity, drug-likeness drug-like nature (according to druglikeness rules: Lipinski, Ghose, Veber, Egan, and Muegge), and

<sup>6</sup> http://lmmd.ecust.edu.cn/admetsar2/

<sup>7</sup> https://www.swissadme.ch

Compound	Ames toxicity	Carcinogens	Acute oral toxicity log (1/(mol/kg))	Rat acute toxicity
CNP0465303	AT	NC	III	2.6938
CNP0455361	NAT	NC	III	2.9773
CNP0457047	NAT	NC	III	2.9038
CNP0139709	NAT	NC	III	3.2080
CNP0458104	NAT	NC	III	2.9703
CNP0478503	NAT	NC	III	2.6235
CNP0475158	NAT	NC	III	2.8388
CNP0447420	AT	NC	III	2.4641
CNP0457847	NAT	NC	III	2.9475
CNP0478727	NAT	NC	III	2.7368
CNP0446191	NAT	NC	III	2.3597
CNP0474861	NAT	NC	III	2.7944
★CNP0348829	NAT	NC	III	2.6326
CNP0475639	NAT	NC	III	2.5313
CNP0475126	NAT	NC	III	2.5391
★CNP0474002	NAT	NC	III	2.4915
CNP0474130	NAT	NC	III	2.7055
CNP0458286	NAT	NC	III	2.3477
CNP0474134	NAT	NC	III	2.4560
CNP0436312	NAT	NC	III	2.3490

#### TABLE 3 Toxicity analysis of the top 20 virtual hits predicted by AdmetSAR2.

Note: AT, Ames toxic; NAT, Non Ames toxic; NC, Non-carcinogenic; Category-III means (500 mg/kg > LD50 < 5,000 mg/kg).

### TABLE 4 The interaction of star hit compounds and X77 with 3CL<sup>pro</sup>.

Compound	Key interaction	No. of NBs	Key residues
CNP0348829	Hydrogen Bond, Pi-Pi Stacked, Pi-Sulfur, Pi-Alkyl	3	HIS163, CYS145, SER144
CNP0474002	Hydrogen Bond, Pi-Alkyl, Halogen, (Fluorine), X-Sulfur, Amid-Pi Stacked	1	CYS145, MET49, GLU166, MET165
Inhibitor (X77)	Hydrogen Bond, Pi-Sulfur, Amid-Pi Stacked, Pi-Alkyl	2	GLU166, LEU141, GLY143, MET49

medicinal chemistry friendliness (PAINS, Brenk, Lead-likeness, Synthetic accessibility) (Daina et al., 2017).

# **Results and discussion**

### Molecular docking results and analysis

With the rapid development of various computer methods such as molecular modeling, molecular docking, and molecular dynamics simulations, modern computer tools provide new strategies to explore the scientific basis of TCM (Jiao et al., 2021). Molecular docking is an established *in silico* structurebased method widely used in drug discovery. Docking enables the identification of novel compounds of therapeutic interest, predicting ligand-target interactions at the atom level, or delineating structure-activity relationships (SAR), without knowing *a priori* the chemical structure of other target modulators (Pinzi and Rastelli, 2019).

It has been shown that structural domain III (residues 201-301) of the 3CL<sup>pro</sup> has catalytic activity (Tahir Ul Qamar et al., 2020). X77 can bind at the active site to inhibit the catalytic activity of the 3CL<sup>pro</sup> (Gahlawat et al., 2020). The interaction diagram of inhibitor X77 bound to 3CL<sup>pro</sup> is shown in Figure 1A. X77 bound to 3CL<sup>pro</sup> by two hydrogen bonds between residues GLY143 and GLU166, Pi-Sulfur interaction with residues MET49, amide-Pi stacked with residues LEU141, and other hydrophobic interactions. Very interestingly, as shown in the middle part in Figure 1A, X77 is docked into the active pocket by an "X" shape.

The top 20 virtual hits with different structural skeletons are listed in Table 1. The binding affinities of all compounds were assessed through LibDock Scores and -CDOCKER energy value. According to the LibDock score results, all compounds had





(A) RMSD of protein backbone, Lig\_fit\_Prot and Lig\_fit\_Lig for CNP0474002-3CL<sup>pro</sup> complex; (B) RMSF of protein C(alpha) and backbone; (C) RMSF of CNP0474002; (D) The 2D structure of CNP0474002.



apparent interactions with the target protein. The top 20 virtual hits showed stronger interactions with  $3\mbox{CL}^{\mbox{\tiny pro}}$  than the original inhibitor X77. Two -CDOCKER interaction energy values of 38.6734 and 35.6722 kcal/mol, compared to that of X77 (-CDOCKER\_ENERGY = 30.4747), were calculated, indicating that CNP0348829 and CNP0474002 are promising candidate compounds for the treatment of SARS-CoV-2. The ADMET property predictions of the virtual hits were demonstrated in Tables 2, 3 in order to investigate how molecules can access the target site of 3CL<sup>pro</sup> after entering the bloodstream. This analysis is also crucial for analyzing the efficacy of molecules (Fadaka et al., 2022). All parameters were within the Lipinski's rule of five (ROF) cut-off range for the test compounds except CNP0465303 and CNP0447420 and present no bystander toxicity effects, since managing toxicity is the main task in developing new medications. Ames toxicity, carcinogenic properties, and rat acute toxicity which are important parameters of drug safety, were predicted in the current investigation. According to the results, most of the compounds showed no Ames toxicity, no carcinogenic properties, and low toxicity, meaning they are safe when used as drugs (Fadaka et al., 2021; 2022). The ideal drug molecules fall under such criteria as molecular weight (<500 g/mol), MlogP values ( $\leq$ 4.5), rotatable bonds ( $\leq$ 10), HB acceptors (<10), and number of HB donors (<5) (Lipinski, 2004). We screened out two star-hits based on ROF, while others were eliminated because of the molecular weight or -CDOCKER\_ENERGY values.

Interaction diagrams of **CNP0348829** and **CNP0474002** bound to 3CL<sup>pro</sup> are shown in Figures 1B, C. As seen in Figure 1B, **CNP0348829** was stabilized by three strong hydrogen bond interactions between residues SER144, CYS145, and HIS163. In addition, the ring exhibited very strong Pi–Pi stacked interactions with HIS41, Pi–sulfur interaction with CYS44. **CNP0474002**-3CL<sup>pro</sup> bound to 3CL<sup>pro</sup> by hydrogen bond interactions between C=O of amide with CYS145. The Pi–alkyl stacked interactions between residues MET165 and 18-phenyl and X–sulfur interaction with MET49 like **CNP0348829** were also observed (Figure 1C; Table 4). Both **CNP0348829** and **CNP0474002** bound to the 3CL<sup>pro</sup> in a spreading conformation, **CNP0474002** was docked into the active pocket by an "X" shape, somewhat like X77.

# Molecular dynamics simulation results and analysis

To verify the results of molecular docking, molecular dynamics simulations were introduced in this study. Molecular dynamics simulations can show the microscopic evolution of the system at the atomic level and visualize the mechanisms and patterns of experimental phenomena, which is a great advantage in studying the mechanism of ligand-receptor interaction (Adcock and McCammon, 2006).

The change in conformation can be analyzed by the value of RMSD, showing whether the equilibrium is reached during the simulation. When the position-shift between the specific frame and the reference frame is 1–3 Å, it can be considered that the position-shift of the specific frame is not much different from the reference frame. It indicates that a larger conformational change occurred during the simulation when the position-shift change is apparent. The backbone RMSD curve represents the RMSD of the protein backbone over time. Lig\_fit\_Prot curve represents the RMSD of the ligand when the protein-ligand complex is first aligned to the protein backbone and then the RMSD of the heavy atom is measured. It is likely that the ligand has shifted away from its initial binding site if the value is significantly larger than the RMSD of the protein. The lig\_fit\_lig curve represents the RMSD of the ligand when the ligand is aligned in its reference conformation which is used to measure the internal fluctuations of the ligand atoms.

The results of molecular dynamics simulations are shown in Figures 2, 3. The major deviation is in RMSD-P, which shows the protein RMSD with a ligand in the active site. The compound CNP0348829 showed initial light fluctuation in RMSD-L, at around 1 Å and 3 Å. However, it showed very strong fluctuations at 52 ns, mainly resulting from the change of conformation at the 12-phenyl, while remaining in the active site of the receptor and stabilized for the remaining period of time in the range of 3.5 Å to 5.5 Å. (Figures 2A, C, D). Unlike CNP0348829, CNP0474002 was stabilized in the range of 3 Å to 5.5 Å at all times under study (Figure 3A). The major initial fluctuation in the 3CL<sup>pro</sup> was mainly due to C-terminal fluctuations, which was confirmed by the RMSF plot for the protein C (alpha) and backbone, where except for C-terminal, all other residues showed RMSF less than 2.5 Å. Over the period of analysis, C-terminal exhibited RMSF up to 4.5 Å (Figures 2B, 3B). RMSF of the ligand with the protein and the ligand with initial conformation was very stable and was between 1.00 and 2.25 Å and 0.25 to 1.25 Å, respectively (Figure 3C). Protein interactions with the ligands were monitored throughout the simulation, and these interactions can be summarized and classified into four types: hydrogen bonds, hydrophobic interactions, ionic bonds, and water bridges. Among all interactions, hydrogen bonds play an important role in the overall dynamics simulation because of their greater influence on compound specificity, metabolic characteristics, and adsorption parameters (Kruth et al., 2005). The following amino acids, such as HIS41, CYS145, and GLU166 in CNP0474002-3CL<sup>pro</sup> complex, and CYS44, ASN142, GLY143, SER144, and GLU166 in CNP0348829-3CL<sup>pro</sup> complex, contribute greatly to the interaction between protein and ligand. These interactions were categorized by type and summarized as shown in Figures 4A, B. The above analysis suggests that the ligand-receptor complexes were maintained well within a relatively stable range with small fluctuations throughout the molecular dynamic simulations, indicating that the

binding modes of **CNP0348829**-3CL<sup>pro</sup> and **CNP0474002**-3CL<sup>pro</sup> complexes are stable during the molecular dynamics simulation.

# Discussion

Molecular docking is investigated as a computational tool for virtually screening thousands of drug-like candidates for their affinity toward a specific molecular target, provided that the structure of the target protein is available. Molecular docking serves as the foundation for the structure-based drug design (SBDD) process. It is a fast, cost-effective, and simplified computational tool for predicting the scores associated with the association of putative ligand molecules and the target protein of interest. The obtained docking score offers a basic result for the discovery of leading compounds among the collection and may be taken into consideration for drug development plans (Wang and Zhu, 2016). When combined with results from experimental research, molecular dockingbased virtual screening may aid in the process of rational medication design and development (Sethi et al., 2019).

With the thorough virtual screening procedure, molecular dynamics simulation also attracted a lot of attention in the search for potential hit compounds. (Ganesan et al., 2017). The most important feature of using molecular dynamics simulation in drug discovery pipelines is that it offers dynamic structural data on interactions between ligands and target receptors with a collection of complexes conformations (Gioia et al., 2017; Saurabh et al., 2020). According to current trends, molecular docking combined with molecular dynamics simulation is being actively studied to confirm potential interactions between ligand and receptor, which helps develop effective drug candidates.

For many decades, ingredients derived from plants have been thought to be the most explored drug candidates. It is possible to assess the potential use of plant-based metabolites with antiviral activity in the treatment of SARS-CoV-2 infections (Bhuiyan et al., 2020; Kumar Verma et al., 2021). In this context, it is crucial to use computational techniques to discover plant-derived phytochemicals that have an inhibitory effect on SARS-CoV-2 infections. For example, Hesperidin, Theaflavin, Biorobin (Peele et al., 2020); Rutin, (Das et al., 2021); Andrographolide (Enmozhi et al., 2021); Asiatic acid, Avicularin Guaijaverin, Chrysanthemin (Amin et al., 2021); and Hypericin (Pitsillou et al., 2020) have been screened as potential therapeutic drugs for 3CL<sup>pro</sup> of SARS-CoV-2 by SBDD, and some of which have had protease inhibition activity demonstrated *in vitro* experiments.

SARS-CoV-3CL<sup>pro</sup> consists of 306 amino acids and functionally inhibits replicase precursor polyproteins; further preventing COVID-19 gene expression and replication. There are three domains in the structure of 3CL<sup>pro</sup>, in which 8-101 residues form domain-I, 102-184 residues form domain-II with six-stranded anti-parallel  $\beta$ -barrel, and 201-303 residues form domain-III, which encompasses five a-helices grouped into the anti-parallel globular cluster and is linked to Domain-II by a long loop region (residues 185-200). Firstly, we identified the active pocket of the 3CL<sup>pro</sup>, which lies in a cleft between Domain-I and II and has a Cys145-His41 catalytic dyad that recognizes P1 Gln and P2 Leu/Met/Phe/Val as substrates, based on the binding site of the known standard inhibitor X77 named N-(4tert-butylphenyl)-N-[(1R)-2-(cyclohexylamino)-2-oxo-1-(yridine-3yl)ethyl]-1H-imidazole-4-carbox amide. In this study, we targeted the SARS-CoV-3CL<sup>pro</sup>; rapidly high-throughput virtual screening was performed to obtain the noncovalent SARS-CoV-2 3CL<sup>pro</sup> inhibitor from a natural product compounds database that includes 407270 natural products. In the subsequent virtual screening, we used X77 as a positive control and screened star-hits with -CDOCKER scores higher than X77. Amino acids including PHE140, GLN189, GLY143, GLU166, THR190, CYS145, and HIS163 were reported to be crucial in the inhibition process of M<sup>pro</sup> through hydrogen bonding, which was also verified in the results of our study (Jin et al., 2020). Gyebi et al. (2021) identified the importance of CYS145 and HIS41 as conserved catalytic dyad residues of SARS-CoV-2 Mpro. Interestingly, compounds CNP0348829 and CNP0474002 interacted with one of the two amino acids by hydrogen bond, and these interactions were also monitored in molecular dynamics simulations for the ligand-receptor complexes. According to the above analysis, we suggest that CNP0348829 and CNP0474002 could be potential inhibitors of 3CL<sup>pro</sup>. Nevertheless, further experiments are needed for activity validation in the future.

# Conclusion

Phytochemical screening *in silico* is especially alluring since it can digitally screen hundreds of compounds in a certain amount of time and examine the possibility of drug-like

# References

Adcock, S. A., and McCammon, J. A. (2006). Molecular dynamics: Survey of methods for simulating the activity of proteins. *Chem. Rev.* 106 (5), 1589–1615. doi:10.1021/cr040426m

Amin, S. A., Ghosh, K., Gayen, S., and Jha, T. (2021). Chemical-informatics approach to COVID-19 drug discovery: Monte Carlo based QSAR, virtual screening and molecular docking study of some in-house molecules as papain-like protease (PLpro) inhibitors. *J. Biomol. Struct. Dyn.* 39 (13), 4764–4773. doi:10.1080/07391102.2020.1780946

Bhuiyan, F. R., Howlader, S., Raihan, T., and Hasan, M. (2020). Plants metabolites: possibility of natural therapeutics against the COVID-19 pandemic. *Front. Med. (Lausanne)* 7, 444. doi:10.3389/fmed.2020.00444

Daina, A., Michielin, O., and Zoete, V. (2017). SwissADME: A free web tool to evaluate pharmacokinetics, drug-likeness and medicinal chemistry friendliness of small molecules. *Sci. Rep.* 7, 42717. doi:10.1038/srep42717

Das, S., Sarmah, S., Lyndem, S., and Singha Roy, A. (2021). An investigation into the identification of potential inhibitors of SARS-CoV-2 main protease using molecular docking study. *J. Biomol. Struct. Dyn.* 39 (9), 3347–3357. doi:10.1080/ 07391102.2020.1763201

Enmozhi, S. K., Raja, K., Sebastine, I., and Joseph, J. (2021). Andrographolide as a potential inhibitor of SARS-CoV-2 main protease: An *in silico* approach. *J. Biomol. Struct. Dyn.* 39 (9), 3092–3098. doi:10.1080/07391102.2020.1760136

molecules. In this study, we performed structure-based virtual screening from a natural products library containing 407270 phytochemicals/traditional Chinese medicinal compounds, based on the crystal structures of the 3CL<sup>pro</sup> as an attractive target for small-molecule oral therapeutics for treating COVID-19. Through layers of virtual screening, **CNP0348829** and **CNP0474002**, as two selected star-hits, are promising as clinical candidate compounds for the treatment of SARS-CoV-2. Further experimental studies are suggested to check the possible preclinical and clinical efficacy of **CNP0348829** and **CNP0474002** for the prevention and treatment of COVID-19.

## Data availability statement

The original contributions presented in the study are included in the article/supplementary material, further inquiries can be directed to the corresponding author.

# Author contributions

All authors listed have made a substantial, direct, and intellectual contribution to the work and approved it for publication.

# Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Fadaka, A. O., Aruleba, R. T., Sibuyi, N. R. S., Klein, A., Madiehe, A. M., and Meyer, M. (2022). Inhibitory potential of repurposed drugs against the SARS-CoV-2 main protease: A computational-aided approach. *J. Biomol. Struct. Dyn.* 40 (8), 3416–3427. doi:10.1080/07391102.2020.1847197

Fadaka, A. O., Sibuyi, N. R. S., Martin, D. R., Klein, A., Madiehe, A., and Meyer, M. (2021). Development of effective therapeutic molecule from natural sources against coronavirus protease. *Int. J. Mol. Sci.* 22 (17), 9431. doi:10.3390/ ijms22179431

Gahlawat, A., Kumar, N., Kumar, R., Sandhu, H., Singh, I. P., Singh, S., et al. (2020). Structure-based virtual screening to discover potential lead molecules for the SARS-CoV-2 main protease. *J. Chem. Inf. Model* 60 (12), 5781–5793. doi:10. 1021/acs.jcim.0c00546

Ganesan, A., Coote, M. L., and Barakat, K. (2017). Molecular dynamics-driven drug discovery: Leaping forward with confidence. *Drug Discov. Today* 22 (2), 249–269. doi:10.1016/j.drudis.2016.11.001

Gioia, D., Bertazzo, M., Recanatini, M., Masetti, M., and Cavalli, A. (2017). Dynamic docking: a paradigm shift in computational drug discovery. *Molecules* 22 (11), 2029. doi:10.3390/molecules22112029

Gyebi, G. A., Ogunro, O. B., Adegunloye, A. P., Ogunyemi, O. M., and Afolabi, S. O. (2021). Potential inhibitors of coronavirus 3-chymotrypsin-like protease (3CLpro): An *in silico* screening of alkaloids and terpenoids from african

medicinal plants. J. Biomol. Struct. Dyn. 39 (9), 3396-3408. doi:10.1080/07391102. 2020.1764868

Huang, K., Zhang, P., Zhang, Z., Youn, J. Y., Wang, C., Zhang, H., et al. (2021). Traditional Chinese Medicine (TCM) in the treatment of covid-19 and other viral infections: efficacies and mechanisms. *Pharmacol. Ther.* 225, 107843. doi:10.1016/j. pharmthera.2021.107843

Jiao, X., Jin, X., Ma, Y., Yang, Y., Li, J., Liang, L., et al. (2021). A comprehensive application: molecular docking and network pharmacology for the prediction of bioactive constituents and elucidation of mechanisms of action in component-based Chinese medicine. *Comput. Biol. Chem.* 90, 107402. doi:10.1016/j. compbiolchem.2020.107402

Jin, Z., Du, X., Xu, Y., Deng, Y., Liu, M., Zhao, Y., et al. (2020). Structure of Mpro from SARS-CoV-2 and discovery of its inhibitors. *Nature* 582 (7811), 289–293. doi:10.1038/s41586-020-2223-y

Kruth, J., Mercelis, P., Van Vaerenbergh, J., Froyen, L., and Rombouts, M. (2005). Binding mechanisms in selective laser sintering and selective laser melting. *Rapid Prototyp. J.* 11 (1), 26–36. doi:10.1108/13552540510573365

Kumar Verma, A., Kumar, V., Singh, S., Goswami, B. C., Camps, I., Sekar, A., et al. (2021). Repurposing potential of Ayurvedic medicinal plants derived active principles against SARS-CoV-2 associated target proteins revealed by molecular docking, molecular dynamics and MM-PBSA studies. *Biomed. Pharmacother.* 137, 111356.

Li, Z. Y., Su, C. Y., and Ding, B. (2019). Molecular dynamics simulation of  $\beta$ -adrenoceptors and their coupled G proteins. *Eur. Rev. Med. Pharmacol. Sci.* 23 (14), 6346–6351. doi:10.26355/eurrev\_201907\_18458

Lipinski, C. A. (2004). Lead- and drug-like compounds: The rule-of-five revolution. *Drug Discov. Today Technol.* 1 (4), 337-341. doi:10.1016/j.ddtec. 2004.11.007

Mandal, A., Jha, A. K., and Hazra, B. (2021). Plant products as inhibitors of coronavirus 3CL protease. *Front. Pharmacol.* 12, 583387. doi:10.3389/fphar.2021. 583387

Mohan, M., James, P., Valsalan, R., and Nazeem, P. A. (2015). Molecular docking studies of phytochemicals from Phyllanthus niruri against Hepatitis B DNA Polymerase. *Bioinformation* 11 (9), 426–431. doi:10.6026/97320630011426

Okimoto, N., Futatsugi, N., Fuji, H., Suenaga, A., Morimoto, G., Yanai, R., et al. (2009). High-performance drug discovery: Computational screening by combining docking and molecular dynamics simulations. *PLoS Comput. Biol.* 5 (10), e1000528. doi:10.1371/journal.pcbi.1000528

Peele, K. A., Potla Durthi, C., Srihansa, T., Krupanidhi, S., Ayyagari, V. S., Babu, D. J., et al. (2020). Molecular docking and dynamic simulations for antiviral compounds against SARS-CoV-2: a computational study. *Inf. Med. Unlocked* 19, 100345. doi:10.1016/j.imu.2020.100345

Pinzi, L., and Rastelli, G. (2019). Molecular docking: shifting paradigms in drug discovery. Int. J. Mol. Sci. 20 (18), 4331. doi:10.3390/ijms20184331

Pitsillou, E., Liang, J., Karagiannis, C., Ververis, K., Darmawan, K. K., Ng, K., et al. (2020). Interaction of small molecules with the SARS-CoV-2 main protease *in silico* and

in vitro validation of potential lead compounds using an enzyme-linked immunosorbent assay. Comput. Biol. Chem. 89, 107408. doi:10.1016/j.compbiolchem.2020.107408

Sahebnasagh, A., Avan, R., Saghafi, F., Mojtahedzadeh, M., Sadremomtaz, A., Arasteh, O., et al. (2020). Pharmacological treatments of COVID-19. *Pharmacol. Rep.* 72 (6), 1446–1478. doi:10.1007/s43440-020-00152-9

Saurabh, S., Sivakumar, P. M., Perumal, V., et al. (2020). "Molecular dynamics simulations in drug discovery and drug delivery," in *Integrative nanomedicine for new therapies* (Berlin, Germany: Springer), 275–301.

Sethi, A., Joshi, K., Sasikala, K., and Alvala, M. (2019). Molecular docking in modern drug discovery: principles and recent applications. *Drug Discov. development-new Adv.* 2, 1–21. doi:10.5772/intechopen.85991

Srivastava, N., Garg, P., Srivastava, P., and Seth, P. K. (2021). A molecular dynamics simulation study of the ACE2 receptor with screened natural inhibitors to identify novel drug candidate against COVID-19. *PeerJ* 9, e11171. doi:10.7717/peerj.11171

Tahir Ul Qamar, M., Alqahtani, S. M., Alamri, M. A., and Chen, L. L. (2020). Structural basis of SARS-CoV-2 3CLpro and anti-COVID-19 drug discovery from medicinal plants. J. Pharm. Anal. 10 (4), 313–319. doi:10.1016/j.jpha.2020.03.009

Ullrich, S., and Nitsche, C. (2020). The SARS-CoV-2 main protease as drug target. Bioorg Med. Chem. Lett. 30 (17), 127377. doi:10.1016/j.bmcl.2020.127377

Unoh, Y., Uehara, S., Nakahara, K., Nobori, H., Yamatsu, Y., Yamamoto, S., et al. (2022). Discovery of S-217622, a noncovalent oral SARS-CoV-2 3CL protease inhibitor clinical candidate for treating COVID-19. *J. Med. Chem.* 65 (9), 6499–6512. doi:10.1021/acs.jmedchem.2c00117

Wang, C., Horby, P. W., Hayden, F. G., and Gao, G. F. (2020). A novel coronavirus outbreak of global health concern. *Lancet* 395 (10223), 470-473. doi:10.1016/S0140-6736(20)30185-9

Wang, G., and Zhu, W. (2016). Molecular docking for drug discovery and development: A widely used approach but far from perfect. *Future Med. Chem.* 8 (14), 1707–1710. doi:10.4155/fmc-2016-0143

WHO Team (2022). WHO expert meeting on evaluation of traditional Chinese medicine in the treatment of COVID-19. Geneva: WHO, 4–17.

World Health Organization (2022). Weekly epidemiological update on COVID-19-28 December 2022. [EB/OL]. Available at: https://www.who.int/publications/m/ item/weekly-epidemiological-update-on-covid-19—14-december-2022 (Accessed December 14, 2022).

Wu, G., Robertson, D. H., Brooks, C. L., 3rd, and Vieth, M. (2003). Detailed analysis of grid-based molecular docking: a case study of CDOCKER-A CHARMm-based MD docking algorithm. *J. Comput. Chem.* 24 (13), 1549–1562. doi:10.1002/jcc.10306

Yang, H., Lou, C., Sun, L., Li, J., Cai, Y., Wang, Z., et al. (2019). admetSAR 2.0: web-service for prediction and optimization of chemical ADMET properties. *Bioinformatics* 35 (6), 1067–1069. doi:10.1093/bioinformatics/bty707

Zhuang, W., Fan, Z., Chu, Y., Wang, H., Yang, Y., Wu, L., et al. (2020). Chinese patent medicines in the treatment of coronavirus disease 2019 (COVID-19) in China. *Front. Pharmacol.* 11, 1066. doi:10.3389/fphar.2020.01066