Original Article

Inhibitory efficacy of 2, 4-diacetylphloroglucinol against SARS-COV-2 proteins: in silico study

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Received: 11 May 2021 / Accepted: 24 November 2021 / Published online: 30 January 2022
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Abstract
The pandemic of Severe Acute Respiratory Syndrome Coronavirus-2 has affected millions of people worldwide with common symptoms of fever, cough, and respiratory complications. The pandemic has posed a huge challenge to emergency health services due to unavailability of potent therapeutic drugs. The proteins associated with the viral pathogenesis has been identified as suitable targets for drug design and warrants effective drug discovery to abate COVID-19. The papain-like protease (PLpro), nucleocapsid (N), main protease (Mpro) and non-structural protein (nsp12) of SARS-CoV-2, key component of processing of viral polyproteins, transcription, assembly and replication. On this streak, present study evaluated the interaction of ligand 2,4-diacetylphloroglucinol (DAPG) with viral proteins using molecular docking with (i) AutoDock 4.2.6 and (ii) AutoDock Vina followed by molecular dynamic simulation studies of protein-ligand complex configuration. The analysis revealed that PLpro (3E9S) and N (4J3K) protein corresponds to the highest docking score and therefore, selected for molecular dynamics simulation study (100 ns). The study comprised analysis of parameters: (i) RMSD and RMSF, (ii) radius of gyration- which indicated interaction of protein entities with ligand supported steadiness of the complex, (iii) Coulombic and Lennard–Jones interactions, which played a significant role in complex stability. DAPG showed a good number of H-bonds with PLpro and MM-PBSA binding energy when compared to the N protein. This study showed DAPG as a potential bioactive molecule to act as an inhibitor for the PLpro thereby, DAPG can be used as potential inhibitor against SARS-CoV-2 and is potential drug candidate against COVID-19.

Keywords SARS-CoV-2 · COVID-19 · DAPG · Molecular docking · Molecular dynamic simulation · MM-PBSA

Introduction
The Coronavirus Disease 2019 (COVID-19) pandemic disease is caused by the novel coronavirus SARS-CoV-2, a genetic homolog of the Severe Acute Respiratory Syndrome

Abbreviations
COVID-19 Coronavirus Disease 2019
SARS Severe Acute Respiratory Syndrome
CoVs Coronaviruses
SARS-CoV-2 Severe Acute Respiratory Syndrome Coronavirus-2
PLpro papain-like protease
N nucleocapsid
Mpro main protease
nsp12 non-structural protein
DAPG 2, 4-diacetylphloroglucinol

RMSD root mean square deviation
RMSF root mean square fluctuation
Rg radius of gyration
MM-PBSA Molecular Mechanics Poisson-Boltzmann Surface Area
ACE-2 Angiotensin Converting Enzyme-2
MD Molecular dynamics
GROMACS Groningen Machine for Chemical Simulations
LJ-SR Lennard–Jones short-range
Coul-SR Coulombic short-range
SASA solvent accessible surface area
GPCR G Protein-Coupled Receptors

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(SARS) coronavirus. The virus is transmitted from human-to-human contact through droplet nuclei during coughing or sneezing. The viral infection has an incubation period of around 5.2 days from the first symptom to 14 days with high incidence to death (Shang et al. 2020). The disease spread rapidly across the globe currently beyond 207 million confirmed infections despite record early approved vaccination drives (WHO 2021).

In the past, the zoonotic human Coronaviruses (CoVs) have caused not only SARS that spread across the globe with mortality rate of 15%, but also MERS-CoV (Middle East Respiratory Syndrome Coronavirus) which recorded a mortality rate of 35% (Chan et al. 2013). The CoVs are grouped into order Nidovirales; family Coronavirus and grouped into four genera: α-CoVs, β-CoVs, γ-CoVs, and δ-CoVs. The SARS-CoV-2 phylogenetically clusters to the β genus, equipped with organisational proteins: (i) spike (S), (ii) envelope (E), (iii) membrane (Mpro), (iv) nucleocapsid (N), (v) papain like protease (PLpro), (vi) RNA dependent RNA polymerase (RdRp), and (vii) RNA binding N terminal domain (NTD) which orchestrate COVID-19 pathogenesis and viral life cycle (Gil et al. 2020). These structural and functional proteins are recognized as suitable drugs against the virus, albeit the genes encoding these proteins are undergoing frequent mutations (Amamuddy et al. 2020) leading to formation of new variants (WHO 2021). During virus infection, spike protein determines the host cell specificity through its glycoprotein and receptor Angiotensin Converting Enzyme-2 (ACE-2) present on epithelial cells of lungs, kidneys, and heart cellular system (Zumla et al. 2016; Kulkarni et al. 2020).

The SARS-CoV-2 genome specifically encodes for two polyproteins, viz. Ppla (450 kD) and Pplb (750 kD), and both are processed with the help of chymotrypsin-fold protease (33 kD) to diverse useful proteins viz. (i) spike, (ii) membrane, (iii) envelop, (iv) nucleoprotein, (v) replicate, and (vi) polymerase. The protease, called main protease (Mpro) is the most suitable target for drug design (Luan et al. 2020; Bello 2020; Sepay et al. 2021). The Mpro is a dimer, comprising two protomers, containing three domains, domains I (residues 8 – 101) and II (residues 102 – 184) that contain β-barrels and domain III (residues 201 – 306) which consists mainly of the α-helices. Drug discovery requires the dimeric state of SARS-CoV-2 Mpro for good results because monomer of Mpro is mostly inactive (Bello 2020; Mengist et al. 2021). Saquinavir and diastereomers of nelfinavir have been identified as a potent inhibitor of dimeric SARS-CoV-2 Mpro (Bello et al. 2020; Sargolzaei 2021).

Presently, various potential broad-spectrum antiviral drugs have been suggested for therapeutic and prophylactic purpose. However, few promising clinically approved beneficial drug/s, specific medicine, inhibitors (especially, targeting proteases), or treatment against SARS-CoV-2 is available. At present, only 8 candidate vaccines are approved, 11 for early or limited use, 127 vaccines are under various stages of human clinical trials and 99 preclinical candidates are under investigation. Despite the global herculean efforts, none of the vaccines warrants to secretary IgA antibody production or protection against the new variants. Several efforts to target structural and functional proteins to: (i) prevent the viral RNA synthesis, (ii) inhibit virus replication, (iii) block the virus binding to ACE-2, and (iv) arrest the virus’s self-assembly process have been reported. Three principal approaches for developing new drugs has emerged (Zumla et al. 2016), namely, (i) testing of known broad-spectrum antiviral compounds (Francés-Monerris et al. 2020), (ii) screening for various drug molecules with therapeutic potential against coronavirus using current molecular databases (Wang 2020), and (iii) advance new targeted drugs with the help of genomic data and pathological characteristics of diverse CoVs (Spreatco et al. 2020). These strategies may prove to be effective, but requires long time span, cost intensive and suffers from high failure rates. Hence, search for potential lead molecules from the marketed drugs appears as the most preferred alternative to manage SARS-CoV-2 infection.

Lonafarnib, tegobuvir, olysio, filibuvir, cepharanthine, theaflavins-3-3′-digallate, epigallocatechin gallate, epicatechingallate, gallocatechin-3-gallate polyphenols, hydroxychloroquine, chloroquine, remdesivir, thiopurine analogs, desacetylgedunin and several phenolic compounds act as protease inhibitors (Ruan et al. 2020; Shivanika et al. 2020; Ghosh et al. 2020; Baildya et al. 2020; Mirza et al. 2020; Baidlya et al. 2021a, 2021b; Sabet et al. 2021). Various antiviral compounds (including FDA approved) as well as experimental drugs have been examined to protect against COVID-19 disease using in silico approach. For this, software tools viz. (i) molecular docking, (ii) molecular dynamic simulation, (iii) target point determination, and (iv) chemical stability appears first line of effective strategy for rapid screening in the emergency coronavirus pandemic situation. In particular, molecular docking rapidly identifies potential new drugs and evaluate preliminary efficacy (Grinter and Zou 2014). In short span, numerous in silico studies has recorded more than 900 antiviral drugs against protein targets involved in viral pathogenesis (mainly proteases), but targeted only a single viral protein entity (Joshi et al. 2020). Preferentially, search for robust inhibitors from natural lead compounds and, chemical inhibitors with capacity to target multiple proteins capable to arrest cross-infection appears promising choice as against single-entity viral targets (Baidlya et al. 2021a, 2021b).

Among the natural compounds, 2,4-diacetylphloroglucinol (DAPG) is a bioactive metabolite with low-molecular weight, non-volatile phenolic polyketide, with myriad environmental and clinical role. DAPG is reported to be
produced by plants, algae, and especially, soil microbes (plant growth promoting rhizobacteria *Pseudomonas* spp.) making it amenable to industrial fermentative production. DAPG has been reported for its antagonistic activity against bacteria, fungi, helminths, protozoa, cancerous cells, and mycobacteria (Kankariya et al. 2019). Notably, DAPG has been reported to display strong activity against enveloped RNA and DNA viruses, such as (i) vesicular stomatitis virus (VSV), an enveloped RNA virus, and (ii) herpes simplex virus (HSV-I), an enveloped DNA virus, but unable to affect non-enveloped RNA virus (polio-I) (Tada et al. 1990).

Biological production of DAPG is usually slow with very low yield, but DAPG can be chemically synthesized, a production route amenable for industrial scale-up (Zakrzewski et al. 2007). Thus, DAPG appears as bespoke bioentity with proven activity against enveloped viruses, fungal and bacterial pathogens which can be economically produced in purified form, tailor made to function, industrial-scale ready to make available in short span to control the COVID-19 pandemic as well as cross infection. Based on this premise, DAPG was selected for virtual screening against papain-like protease (PLpro), nucleocapsid (N), main protease (Mpro) and non-structural protein (nsp12) of SAR-CoV-2. The PDB: 3E9S – papain like protease has role in cleavage and maturation of viral polyproteins, assembly of the replicase-transcriptase complex, interruption of host responses, transmission and virulence of the virus (Osipiuk et al. 2021; Sabet et al. 2021). PDB: 4J3K and PDB: 1SSK- nucleocapsid (N) – RNA-binding protein dangerous for viral genome packaging and virus particle release (Zeng et al. 2020; Cubuk et al. 2021). PDB: 6LU7- main protease (Mpro) plays a critical role in viral replication and maturation (Joshi et al. 2020; Mengist et al. 2021). PDB: 6NUR- non-structural protein (nsp12) - essential subunit of RdRp (RNA-dependent RNA polymerase) and participate in RNA package and virus particle release (Ruan et al. 2020; Zeng et al. 2020).

In the present study, Molecular docking study has been performed and higher binding energy was observed, which is being reported for the first time. Also we report the molecular dynamics simulation to assess the stability of proteins (of two ligand protein complexes which exhibited strong stability) with the DAPG ligand by estimating diverse parameters like SASA (solvent accessible surface area), RMSF (root mean square fluctuation), RMSD (root mean square deviation), and Rg (radius of gyration analysis). Prediction of Activity Spectra for Substances (PASS) and Bioactivity score analysis also depicted the drug-like nature of ligand and complexes. Thus further validation of DAPG and PLpro of SARS-CoV-2 through wet lab and clinical studies is warranted.

### Materials and methods

#### Molecular docking analysis

**AutoDock 4.2.6**

The crystal structures of SARS-CoV-2 proteins were downloaded from RCSB Protein Data Bank (PDB) database as follows: (i) PDB ID: 3E9S (papain-like protease) (Ratia et al. 2008), (ii) PDB ID: 4J3K (nucleocapsid N protein) (Chen et al. 2013), (iii) PDB ID: 6LU7 (SARS-CoV-2 main protease) (Jin et al. 2020), (iv) PDB ID: 1SSK (nucleocapsid protein) (Huang et al. 2004), and (v) PDB ID: 6NUR (SARS-CoV NSP12) (Kirchdoerfer and Ward 2019).

The structure of the DAPG was retrieved from PubChem database. All the molecules from PDB and PubChem databases were converted to *.pdbqt format using Autodock tool and used for initial docking studies. The molecular docking technique examined for the folding pattern and molecular interactions between proteins and DAPG ligand molecule. Molecular docking was performed by AutoDock 4.2.6 program, using the implemented empirical free energy function and the Lamarckian Genetic Algorithm (LGA) for ligand tethering of proteins (Morris et al. 2009; Joshi et al. 2020; Shivanika et al. 2020). The interactions of complex protein-ligand conformation including hydrogen bonds and bond lengths were analyzed using Pymol software, UCSF Chimera, Molegro Molecular Viewer and Accelrys DS Visualizer software. Further validation through additional docking software was performed using AutoDock Vina.

**AutoDock Vina**

The docking of viral proteins with the DAPG was performed with the aid of AutoDock Vina, which is now widely used as a successor of AutoDock Tools (Ghosh et al. 2020). The binding affinities of DAPG–viral proteins were determined and analyzed using the same software. Execution of AutoDock Vina is faster than AutoDock 4.2.6, which reduces the size of the conformational space, allowing it to be searched reliably and reduces the computational effort in predictions of binding pockets. AutoDock Vina was used to perform docking simulations, generating conformations of ligand in complex with the receptor, which were finally ranked on the basis of binding energy (Siddiqui et al. 2020).

#### Molecular dynamics (MD) simulation study

MD simulation (100 ns) was achieved with the minimum energy conformer of two proteins and DAPG (PLpro-DAPG and N-DAPG) complex using Groningen Machine for
Chemical Simulations (GROMACS) version 2020.2 (Krupanidhi et al. 2020). For the simulation, the protein topology was generated using the pdb2gmx tool of GROMACS. The ligand topology was generated using ANTECHAMBER from Amber Tools 16. The amber ff99SB-ILDN force field was chosen with the TIP3P water model. Amber ff99sb-ILDN (Lindorff-Larsen et al. 2010) is designed for protein, improved version by refitting the amino acid side chain torsion potentials of the AMBER99SB force field for four residues: isoleucine, leucine, aspartic acid, and asparagine (Aliev et al. 2014). Also, it has been used in several protein dynamics and found to give better results (Fan et al. 2021). The ligand topology generated using ANTECHAMBER from AmberTools 16 wherein we have used amber ff99sb charge model. The complex was placed in the box and was defined as a cubic unit cell with 1 nm. For PLpro-DAPG and N-DAPG complex, the system was soluted by adding 44,395 and 9655 water molecules and neutralized by adding 3 and 5 sodium ions, respectively. The energy minimization of each system was performed using the steepest descent method for force <10.0 kJ mol$^{-1}$. Solvents and ions around the target proteins were equilibrated, maintaining temperature (300 K) and target pressure (1 bar). The produced trajectories were used to calculate root mean square deviation (RMSD), root mean square fluctuation (RMSF), radius of gyration (Rg), and hydrogen bond graphs by using numerous in-built scripts of GROMACS.

**Binding free energy calculation**

It was calculated by the Molecular Mechanics Poisson-Boltzmann Surface Area (MM-PBSA) method. For this purpose, binding free energy was calculated by the ‘g_mmpbsa’, a GROMACS tool and the tree-step calculation was performed to calculate interaction energy between enzyme and ligand (Kumari et al. 2014). The potential free energy in vacuum was calculated for polar solvation and non-polar solvation energy using SASA model. Python script MmPbSaStat.py, provided in the g_mmpbsa package, was used to calculate the binding energy components of protein-ligand binding. The binding free energy was calculated using,

$$
\Delta G_{\text{bind}} = G_{\text{complex}} - (G_{\text{protein}} + G_{\text{ligand}})
$$

where, $\Delta G_{\text{bind}}$ represents the binding free energy for complex (protein and ligand), $G_{\text{complex}}$, $G_{\text{protein}}$ and $G_{\text{ligand}}$ are the absolute free energy for complex, protein, and ligand, respectively.

**Prediction of activity spectra for substances (PASS) analysis**

It predicts biological activity spectrum of a compound under study based on its structure-activity relationship with a known chemical entity. PASS analysis was performed using various online tools as detailed viz. Lipinski’s rule of five. Lipinski’s rule of five is helpful in describing molecular properties of drug compounds required for estimation of important pharmacokinetic parameters (Siddiqui et al. 2020).

**Bioactivity score prediction (BAS)**

BAS values suggest a compound’s overall ability to be a potent drug candidate. Molinspiration is a web-based online tool which was used to predict the drug scores of the prospective DAPG with respect to several human receptors like GPCRs, ion channels, kinases, nuclear receptors, proteases and enzymes.

**Pharmacokinetic (PK) parameters prediction**

ADMET (A-Absorption, D-Distribution, M-Metabolism, E-Excretion and T-Toxicity) properties of DAPG were predicted using online Swiss ADME software. This software examines the essential pharmacokinetic properties of a compound like distribution viz. blood-brain barrier (BBB) and skin permeability (LogKp), and its metabolism in terms of it being a P glycoprotein (P-gp) substrate, Cytochrome P450 viz. CYP1A2, CYP2C19, CYP2C9, CYP2D6, CYP3A4 inhibitor as well as its lipophilicity for plasma membrane absorption (Siddiqui et al. 2020).

**Results and discussion**

**Molecular docking**

To study the binding interaction of the DAPG compound with the viral proteins (3E9S, 4J3K, 6LU7, 1SSK and 6NUR) molecular docking were performed with the help of AutoDock 4.2.6 program. Based on the binding energy of ligand-protein interactions with AutoDock 4.2.6 tool and ligand-protein complexes were selected for further validation through additional docking tool- AutoDock Vina. AutoDock 4.2.6 analyses revealed that the binding affinities of the DAPG with SARS-CoV-2 proteins which decreased in the order N (4J3K) > PLpro (3E9S) > Mpro (6LU7) > N (1SSK) > NSP12 (6NUR). Table 1 summarizes the binding energy parameters and interacting amino acid residues participating in binding pocket of proteins of SARS-CoV-2 with DAPG.

Results of molecular docking analysis using AutoDock 4.2.6 showed that DAPG has the best binding affinity with nucleocapsid (N) protein [$-5.27$ Kcal mol$^{-1}$] followed by papain like protease (3E9S) [$-5.20$ Kcal mol$^{-1}$]. On the other hand, DAPG exhibited the best binding affinity with
papain like protease (3E9S) and main protease (6LU7) \([-6.20; -5.97 \text{ Kcal mol}^{-1}\) respectively] as analyzed by AutoDock Vina. Intrestingly, this study is in agreement with a previously published report where phytoconstituents from miracle herb \textit{Nigella sativa} which showed similar kind of variation in binding energy with their selected protein targets (Siddiqui et al. 2020). Results propose that slight dissimilarities in binding affinity of the DAPG is because of the variances in the generation of grid boxes and determination of binding pockets on the target proteins by these software on account of slight differences in selection criteria. This has led to a difference in interacting amino acid residues in the binding pockets of proteins of SARS-CoV-2 as is evident from Table 1.

After exploring the molecular interactions, it was perceived that the bioactive molecule DAPG forms various interaction as in case of: (i) papain-like protease (3E9S), manifests hydrogen bonds formed with amino acid residues Glu168, Lys158, Leu163, Gln270, Asp165, and one Pi-Pi T shaped interaction with Gly164. DAPG also formed van der Waals (VdW) interactions with residues Tyr269, Tyr274, and Tyr265; (ii) the Nucleocapsid protein (4J3K) revealed hydrogen bonds with amino acid residues Val80, Ala77, Val74, and few residues formed other interactions like Ala85 (Pi-sigma), Ser138 (C-H), and Ala82 (Pi-alkyl), while residues Pro81, Pro75, Pro165, Ile76 and Tyr115 showed VdW interactions; (iii) the main protease (6 LU7) formed hydrogen bond with amino acid residues Glu166, His164, Arg188, Met165, His41 and other interaction with Met (Pi-alkyl) along with VdW interactions formed with residues Leu167, Gln192, Val186, Gln189, Thr190, Met49 and Asp187; (iv) Nucleocapsid protein (1SSK) residues, the Trp30, Phe88, Gly125, and Thr27 formed hydrogen bonds, while several residues such as Ser29, Phe31,Tyr87, Tyr89, Arg66 formed VdW interactions; (v) the NSP12 (6NUR), residues viz. Asp517, Gln524, Tyr521, Val495 formed hydrogen bonds, Leu498 formed Pi-alkyl interaction and VdW interactions formed with Ile494, Val493, Asn491, Tyr516, Asn496 stabilizing the binding of the ligand to the enzyme. Cumulatively, the hydrogen bonding interaction of DAPG ligand with viral proteins was found highly stable, while the prevalence of many hydrogen bonds indicated effective inhibition and DAPG can fit inside the pocket present in the viral protein’s domain (Fig. 1 and Fig. S1). The docking results obtained from AutoDock 4.2.6 and AutoDock Vina tools were visualized in Discovery Studio showing prominent interactions between various amino acid residues (Table 1) and Fig. S2 displayed the best docking poses of DAPG with proteins of SARS-CoV-2.

The results are in accordance with Udrea et al. 2020; Shivanika et al. 2020 and Sabet et al. 2021 who proposed that natural antiviral phytocompounds, natural compounds and phenolic acids could act as a lead for the COVID-19 treatment by using AutoDock 4.2.6. Thioridazine and sulphoridazine compounds are the class of Phenothiazines which displayed binding energy of \(-7.06 \text{ kcal mol}^{-1}\) and \(-8.72 \text{ kcal mol}^{-1}\) (Udrea et al. 2020). Eight green tea polyphenols exhibited good binding affinity \(-7.1 \text{ to } -9.0 \text{ kcal mol}^{-1}\) (Ghosh et al. 2020). Also, Uercetin, Kaempherol, Nelfinavir and Lopinavir has demonstrated binding energy of \(-6.6, -6.4, -7.0 \text{ and } -6.4 \text{ kcal mol}^{-1}\), respectively towards Mpro (Huynh et al. 2020). Similarly, the compounds ADM_13083841, LMG_15521745 and SYN_15517940 with binding energy score \(-8.9, -8.7, -8.7 \text{ kcal mol}^{-1}\), respectively along with 26 FDA-approved drugs revealed a docking score of \(-7\) or higher (Alamri et al. 2020). On a similar streak, anti-malarial drugs, Chloroquine and Hydroxychloroquine, was recognized as an inhibitor

| Table 1 Docking interaction scores of DAPG with various proteins of SARS-CoV-2 |
| --- |
| Ligand | Proteins | AutoDock 4.2.6 | AutoDock Vina |
| --- | --- | --- | --- |
| PLpro (3E9S) | MF: C_10H_10O_5, MW: 210.18 | Docking Score (kcal mol\(^{-1}\)) -5.20 | Docking Score (kcal mol\(^{-1}\)) -6.20 |
| Interacting amino acids | GU168(H1), LYS158(H2), LEU158(H3), GLN270(H4), ASP165(H5), GLY164 | TYR:274(H1), TYR:265(H2), TYR:269, GLN:270, GLY164 |
| N (4J3K) | Docking Score (kcal mol\(^{-1}\)) -5.27 | Docking Score (kcal mol\(^{-1}\)) -5.97 |
| Interacting amino acids | VAL80(H1),ALA77(H2),VAL74, SER138 | THR:27(H1), TRP:30(H2), TYR:90 |
| Mpro (6LU7) | Docking Score (kcal mol\(^{-1}\)) -4.94 | Docking Score (kcal mol\(^{-1}\)) -5.97 |
| Interacting amino acids | GLU166(H1), HIS164(H2), ARG188, MET165, HIS41 | GLU:166(H1), HIS:163(H2), SER:144(H3), GLY:143(H4) |
| N (1SSK) | Docking Score (kcal mol\(^{-1}\)) -4.71 | Docking Score (kcal mol\(^{-1}\)) -5.80 |
| Interacting amino acids | TRP30(H1), PHE88(H2), GLY125, THR27 | TYR:90, THR:27(H1), TRP:30(H2) |
| NSP12 (6NUR) | Docking Score (kcal mol\(^{-1}\)) -4.61 | Docking Score (kcal mol\(^{-1}\)) -5.97 |
| Interacting amino acids | ASP517(H1), GLN524(H2), TYR521(H3), VAL495(H4) | TYR:456(H1), THR:556(H2), ARG:624(H3) |
| PDB ID 3E9S | PDB ID 4J3K | PDB ID 6LU7 | PDB ID 1SSK |
|-------------|-------------|-------------|-------------|
| AutoDock 4.2.6 | AutoDock Vina | AutoDock 4.2.6 | AutoDock Vina |

**Fig. 1** 2D interactions of SARS-CoV-2 target viral proteins with the DAPG ligand molecule
towards SARS-CoV-2 PLpro (Baildya et al. 2020, 2021a, 2021b; Hussein and Elkhair 2021). Negative binding affinity indicated stronger protein-ligand complex (Kodchakorn et al. 2020). Similarly, Nilotinib, Saquinavir, Tipranavir, Lonafarnib, Tegobuvir, Olysio, Filibuvir, and Cepharanthine drugs are reported for the inhibition of NSP12 protein with binding energy of −7.7 to −8.6 kcal mol\(^{-1}\) (Ruan et al. 2020) by using AutoDock Vina. Interestingly, DAPG showed strong binding with all selected SARS-CoV-2 proteins and is previously reported as an antiviral compound for enveloped viruses (Tada et al. 1990), hence, it may be a potent inhibitor of the targeted key viral proteins used in this study.

**Molecular dynamics (MD) simulation study**

During the 100 ns MD simulations, in PLpro-DAPG complex, the RMSD values obtained between the ranges of 0 to 0.6 nm (Fig. 2), indicating that the protein fold fluctuated during the initial period of the simulation. After 40 ns, the structure started to find an equilibrated position, and just before 50 ns, the complex became dynamically stable. After 50 ns, the protein complex remained stable with 0.1 rms value, indicating that the complex established stability at 50 ns and remained stable for the rest of the span, i.e., the protein interacted with ligand quite well. In case of N-DAPG complex, the RMSD values remained between the ranges of 0 to 0.5 nm, which indicated that the protein complex fold was steady throughout the simulation. After 30 ns, the structure started to find an equilibrated position, and just before 50 ns, the complex became dynamically stable thereafter.

Thus, both PLpro-DAPG and N-DAPG complex remained stable with the average RMSD of 0.1 rms, suggesting that both the complexes established stability at 50 ns and 30 ns, respectively. Also, the complexes remained stable for the rest of span demonstrating faithful protein interaction with DAPG ligand. Although the RMSD of compound fluctuated in the initial period of the run, the complexes stabilized rapidly thereby, displayed negligible RMSD deviation. After 40 ns and 50 ns, for both PLpro-DAPG and N-DAPG complex changes in ligand was observed to be 0.1 nm indicating transformation of binding orientation over the simulation.

![Fig. 2](image-url) RMSD of protein and ligand in PLpro-DAPG complex (a) and (c); and N-DAPG complex (b) and (d)
time. The results indicated that both the complex exhibited RMSD of <0.6 nm.

Further, to measure the fluctuation measure of atoms or residue in simulation, RMSF calculation on residue level was performed. The PLpro-DAPG complex residues at 50, 107, 192, 227, 228 and 281 showed main fluctuations but other residues showed variation in rms value of about 0.3 nm. In case of N-DAPG complex, residues at 70, 95, 98, 100, 101, 106, 123 and 158 showed minor fluctuations, between 95 to 107 showed major fluctuations, while other residues showed variation in rms value of <0.3 nm (Fig. 3).

During MD simulation, ligand steadiness and fluctuation were supported by several hydrogen bonds. In this study, both PLpro-DAPG and N-DAPG complex showed maximum of 4 to 5 hydrogen bonds formed during the simulation along with pairs within 0.35 nm (3.5 Å) which signifies the number of contacts formed within the cut-off value in between protein and ligand. It showed that a maximum of 13 and 11 contacts were formed by DAPG during simulation in PLpro-DAPG and N-DAPG complex, respectively. Most of the frames were found to have 3 to 6 contacts (Fig. 4), suggesting similar conformational stability and such network possibly played a significant role in strengthening the binding effect throughout the simulation period ultimately supporting the docking. Poor stability of ligand-binding is often identified from fewer number of hydrogen bonds, while flexibility alters protein shape, structure, and function. Proteins generally require significant degree of flexibility at physiological conditions to carry out native function. Under these circumstances, DAPG can certainly alter the flexibility of protein and decrease the enzymatic activity.

The radius of gyration (Rg) of a protein complex is a measure of its compactness and stably folded protein likely maintains a relatively steady value of Rg. However, Rg changes over time when protein unfolds (Alamri et al. 2020). In the present MD simulation run, Rg showed some fluctuations in the initial phase, which corroborated to the behaviour observed in the RMSD estimations for both complex, and stabilised under 2.4 nm and 1.5 nm, respectively. The Fig. 5 demonstrates the steadiness of PLpro-DAPG complex after 30 ns and N-DAPG complex after 50 ns.

Two types of short-range interaction energies were calculated: (i) LJ-SR (Lennard–Jones short-range), and (ii) Coul-SR (Coulombic short-range). The average Coul-SR interaction energy for PLpro-DAPG and N-DAPG complex showed to be $-21.87 \pm 1.9$ and $-18.26 \pm 1.7$ kJ mol$^{-1}$, respectively [Fig. 6a], while the LJ-SR interaction energy was $-125.12 \pm 3.1$ and $-87.11 \pm 4.4$ kJ mol$^{-1}$, respectively [Fig. 6b]. Considering the result of 100 ns simulation, contribution of Coul-SR potential in both complexes was greater than LJ-SR potential. The amount of LJ-SR and Coul-SR potential generally contribute to the interaction energy estimation.

**Binding free energy calculation**

To understand the interaction between proteins and DAPG moiety binding energy calculations by MM-PBSA method was performed (Table 2) and binding energies followed the order of PLpro > N. There were strong van der Waal and electrostatic interactions between PLpro and DAPG compared to N and DAPG.

MM-PBSA methodology has developed a popular method for estimated the free energy of a complex (Kollman et al. 2000; Kuhn et al. 2005). The DAPG ligand and protein interactions shown in the form of binding energy was considerably good because lesser the binding energy better is the protein-ligand interaction (Joshi et al. 2020). An aggregate sum of all energy estimates i.e., VdW energy, Electrostatic energy, Polar solvation energy and, SASA energy
was calculated. Relatively, MM-PBSA study showed that PLpro-DAPG complex showed a least binding free energy of \(-96.65 \text{ kJ mol}^{-1}\) and hydrogen bond formation supporting results of docking, hence, it can be claimed that DAPG can inhibit the viral protein PLpro, the Papain like protease.

Similarly, N-DAPG complex showed slightly greater binding free energy of \(-62.58 \text{ kJ mol}^{-1}\). Thus, the study demonstrates a promising stability of DAPG with the viral papain like protease (PLpro) over the simulated MD time of 100 ns. The entropy change upon binding was found to be negligible.
and hence, it can be ignored if relative binding free energy is considered (Wichapong et al. 2016). In the recent past, it was found that naphthalene-based inhibitors have inhibitory effects towards PLpro of SARS-CoV-2, wherein MM-PBSA binding energy from simulations of the four different starting poses (A–D) was between −12.3 to −17.7 kcal mol$^{-1}$. On this streak, a compound, Z93 revealed binding energy of −6.66 kcal mol$^{-1}$ determined through MM-PBSA, the first chemical compound demonstrated against SARS-CoV-2 PLpro (Mirza et al. 2020). Similarly, Phenformin, Quercetin, and Ritonavir showed binding energy values of −56.6, −40.9, and −37.6 kcal mol$^{-1}$ as calculated by MM-GBSA (Molecular Mechanics/Generalized Born Surface Area), while compounds ADM_13083841, LMG_15521745 and SYN_15517940 showed binding energy of 11.89, −10.84, and −12.96 kcal mol$^{-1}$, respectively from the MM-PBSA which demonstrated specific inhibition towards PLpro (Alamri et al. 2020; Kandeel et al. 2020). Recently, Pitsillou et al. 2021 found that Hypericin, Rutin and Cyanidin-3-O-glucoside shows concentration-dependent inhibition of the

Table 2 Different types of interaction energies (kcal mol$^{-1}$ ± SD) between protein and DAPG complex

| Complex      | Van der Waal (VdW) energy | Electrostatic energy | Polar solvation energy | SASA energy | Binding energy |
|--------------|----------------------------|----------------------|------------------------|-------------|---------------|
| PLpro-DAPG   | −31.60                     | −3.65                | 14.75                  | −2.58       | −23.09        |
| N-DAPG       | −21.92                     | −2.38                | 11.41                  | −2.06       | −14.85        |
PLpro in silico as well as in vitro enzymatic inhibition assay. The natural compound, desacetylgedunin extracted from Azadirachta indica (neem) was shown to have highest docking score with PLpro and induce a large structural change as evaluated by MD simulation (Baidyia et al. 2021a, 2021b).

**PASS analysis of DAPG using Lipinski’s rule of five**

Lipinski’s rule of five evaluates drug likeliness of a compound having properties that would make it potential drug in humans. Table 3 shows the PASS analysis of DAPG in terms of their physicochemical properties by applying Lipinski’s rule of five. The oral activity of a drug compound is predicted by calculating the parameters like partition coefficient, polar surface area, number of hydrogen bond donors, number of hydrogen bond acceptors, and molecular weight. An orally active drug compound has no more than one violation (Khan et al. 2017). When compared with hydroxychloroquin, DAPG also showed Lipinski’s violation 0 (Siddiqui et al. 2020).

**Bioactivity score (BAS) prediction**

The predicted BAS of DAPG summarized in Table 4. Normally, a molecule having BAS >0.00 is most likely to possess considerable biological activities, while compounds having values between −0.50 and 0.00 are supposed to be moderately active and compounds having BAS < -0.50, are expected to be inactive. The results of the present study demonstrated that DAPG is biologically active molecule and capable of producing the physiological actions by multiple mechanisms after interacting with G Protein-Coupled Receptors (GPCRs) ligands, nuclear receptor ligands or by acting as inhibitors of proteases and other enzymes. DAPG displayed considerable activity as protease inhibitor as evident from their BAS score.

**Pharmacokinetic (PK) parameters prediction**

To assess the pharmacokinetic feasibility of DAPG as prospective drug candidate, its ADMET properties viz. absorption, distribution, metabolism, excretion and toxicity were calculated using online SwissADME software (Table 5).

Based on the calculated Log P value, DAPG found to be lipid soluble. Intriguingly, DAPG neither displayed blood-brain barrier (BBB) permeability nor predicted to act as permeability-glycoprotein (P-gp) substrate. The CYPs (Cytochromes P450) are involved in the biotransformation of xenobiotics. Compounds that inhibit the five classes of CYPs viz. CYP3A4, CYP1A2, CYP2C9, CYP2C19 and CYP2D6 would cause an increase in their plasma concentrations, thus contributing to improved bioavailability.

### Table 3 PASS analysis of DAPG [Lipinski’s rule of 5 (Physicochemical properties)]

| Ligand                  | MW (<500) | Topological Polar Surface Area (Å²) (TPSA) (<160 Å²)² | Heavy atom count (n atoms) | Hydrogen Bond Donors (nOHNH) (≤5) | Hydrogen Bond Acceptors (nON) (≤10) | Number of Rotatable bonds (≤10) | Lipinski’s violation |
|-------------------------|-----------|------------------------------------------------------|-----------------------------|----------------------------------|-------------------------------------|-------------------------------|---------------------|
| 2,4-diacetylphloroglucinol | 210.18    | 94.83                                                | 15                          | 3                                | 5                                   | 2                             | 0                   |

### Table 4 Bioactivity score of DAPG

| Ligand                  | GPCR ligand | Ion channel modulator | Kinase inhibitor | Nuclear receptor ligand | Protease inhibitor | Enzyme inhibitor |
|-------------------------|-------------|-----------------------|------------------|------------------------|-------------------|------------------|
| 2,4-Diacetylphloroglucinol | −0.55       | −0.04                 | −0.67            | −0.35                  | −0.62             | −0.11            |

### Table 5 ADMET properties calculated for DAPG

| Ligand                  | Lipophilicity (Consensus Log Po/w) | BBB permeant | P-gp substrate | CYP1A2 inhibitor | CYP2C19 inhibitor | CYP2C9 inhibitor | CYP2D6 inhibitor | CYP3A4 inhibitor | Log Kp (skin permeation) |
|-------------------------|-----------------------------------|--------------|----------------|------------------|------------------|------------------|------------------|------------------|--------------------------|
| 2,4 diacetylphloroglucinol | 0.83                              | No           | No             | No               | No               | No               | No               | No               | −6.64 cm s⁻¹              |
the present study, DAPG was not found to act as inhibitor of the class of CYPs. Kp is skin permeability, widely used to quantitatively describe the rate of chemical permeation through the outermost layer of the skin. DAPG showed negative Kp value which indicates less possibility of topical absorption of DAPG.

Overall, the concerted approach of molecular docking, MD simulation and interaction energy estimation have been proved to be useful for design and development of potent viral protein inhibitors. Mostly, inhibition of papain-like protease (PLpro) can potentially block continuation of the viral life cycle, viral replication and processing viral polyproteins during maturation. Hence, PLpro signifies an important antiviral target protease using DAPG as an inhibitory compound.

Consequently, this study primarily draws comprehensive merit from the natural bioactive compound DAPG as protease inhibitor to help design more effective inhibitor for the SARS-CoV-2 infection. Remarkably, DAPG has previously shown anti-bacterial, anti-fungal, anti-viral, anti-cancer, anti-helmintic and antiprotozoal activity as a natural anti-biotic and therapeutic agent and hence, DAPG deserves to be designated as a multi-functional natural metabolite.

Further, the chemical derivation of DAPG functional groups is another potential area to extend its potency against the viral proteins using organic approaches. The derivatized-DAPG and viral protein interaction could be further enhanced, evaluated, and can help select even better synthetic lead drug targets, thereby help to design even more potent viral protein inhibitors, suggesting a pandora box for exploring even more effective antiviral drugs from natural sources with negligible side effects.

Conclusion

The present study computationally investigated DAPG compound for prediction of their potential inhibitory activities against papain-like protease (PLpro), nucleocapsid (N), main protease (Mpro) and non-structural protein (nsp12) of SARS-CoV-2. Analysis of favourable docked conformations of DAPG compound revealed the highest binding affinity with PLpro and N protein. MD simulation and evaluation of parameters like RMSD, RMSF and residue analysis of protein revealed that DAPG has a profound effect on the conformational alteration with these two proteins with noticeable highlighting effect especially on PLpro. This is probably the first report which explores the binding potential of DAPG against SARS-CoV-2 and in a multi-target approach. The interaction of DAPG could be further enhanced via functional group derivatizations and designing of novel compounds. Finally, DAPG could serve as potential lead molecule for multi-targeted drug development regime and future clinical studies.

Supplementary information The online version contains supplementary material available at https://doi.org/10.1007/s11756-021-00979-4.

Acknowledgments Dr. Vaibhav Sable, MrBiologist, Pune for significant support in bioinformatic analysis and valuable suggestions.

Author contribution Raksha A. Kankariya: Conceptualization, Methodology, Software, Writing- Original draft preparation; Ambalal B. Chaudhari: Supervision, Writing- Reviewing and Editing; Navin D. Dandi: Conceptualization, Methodology, Software, Supervision, Writing- Original draft preparation, Writing- Reviewing and Editing; Funding acquisition: This work was supported by Department of Science and Technology, New Delhi, India [SR/WOS-A/LS-1209/2014(G); WoS-A scheme to RAK] and [SR/FST/LS-II/2017/101(C); DST-FIST]; University Grants Commission, New Delhi, India [F.4-23/2015/DRS-III (SAP-II); UGC-SAP-DRS (Phase-III)].

Declarations

Conflict of interest The authors declare that they have no conflict of interest.

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