

Original Article

Effective Binding Affinity of Inhibitor N-(3-(Carbamoylamino) Phenyl) Acetamide against the SARS-CoV-2 NSP13 Helicase

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ABSTRACT

Background and objectives: The outbreak of coronavirus disease 2019 (COVID-19) has become a global health emergency. The severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) NSP13 helicase plays an important role in SARS-CoV-2 replication and could serve as a target for the development of antivirals. The objective of the study was to perform homology modeling and docking analysis of SARS-CoV-2 NSP13 helicase as a drug target.

Methods: The structure and function of SARS-CoV-2 NSP13 helicase were predicted by in-silico modeling studies. The SWISS-MODEL structure assessment tool was used for homology modeling and visual analysis of the crystal structure of the protein. The validation for structure models was performed using PROCHECK. Model quality was estimated based on the QMEAN and ProSA. The MCULE-1-Click docking and InterEvDock-2.0 server were used for protein-ligand docking.

Results: The SARS-CoV-2 NSP13 helicase model corresponded to probability confirmation with 90.9% residue of the core section, which highlights the accuracy of the predicted model. ProSA Z-score of -9.17 indicated the good quality of the model. Inhibitor N-(3-(carbamoylamino) phenyl) acetamide exhibited effective binding affinity against the NSP13 helicase. The docking results revealed that Lys-146, Leu-147, Ile-151, Tyr-185, Lys-195, Tyr-224, Val-226, Leu-227, Ser-229 residues exhibit good binding interactions with inhibitor ligand N-(3-(carbamoyl amino) phenyl) acetamide.

Conclusion: Hence, the proposed inhibitor could potently inhibit SARS-CoV-2 NSP13 helicase, which is thought to play key roles during viral replication. The results of this study indicate that N-(3-(carbamoylamino) phenyl) acetamide could be a valuable lead molecule with great potential for SARS-CoV-2 NSP13 helicase inhibition.

Keywords: <u>SARS-CoV-2</u>, <u>COVID-19</u> Testing, <u>DNA</u> <u>Helicases</u>.

INTRODUCTION

The severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) is the causative agent of the ongoing coronavirus disease 2019 (COVID-19) pandemic $(\underline{1}, \underline{2})$. The novel coronavirus was first reported from Wuhan, Hubei Province of China in late 2019 (3-5). The virus has spread worldwide within a year $(\underline{6}-\underline{8})$. The SARS-CoV-2 is a member of the family Coronaviridae (9). Coronaviruses are enveloped, single-stranded RNA viruses with genome sizes ranging from 26 to 32 kb (10-12) and at least 6 (14 in case of SARS-CoV-2) open reading frames (13, 14). The first genome sequence of SARS-CoV-2 deposited in the GenBank was reported from Wuhan-Hu-1 (MN908947) as a ~30 kb isolate (15), which was used for sequence-related analyses in this article. The ongoing outbreak of SARS-CoV-2 has caused tremendous economic and human losses around the world, while there is still no approved drug available for this infection. The current COVID-19 pandemic has prompted worldwide efforts for the rapid identification and development of vaccines and effective antiviral treatments (16, 17). Most previous studies on anti-coronavirus targets have been focused on the main protease, spike protein (S), RNA-dependent RNA polymerase (RdRp, NSP12), NTPase/ NSP13 helicase, and papainlike protease (PLpro, part of NSP3) (18-20). The SARS-CoV-2 NSP13 possesses NTPase and RNA helicase activities and is conserved (21). The NTPase and helicase activities of SARS-CoV-2 NSP13 may play an important role in SARS-CoV-2 replication and could serve as a target for antivirals (20). In addition, NSP13 ZBD conformations indicate the role of induced-fit flexibility at the ligand binding site (40). Drug hits have significant activity in inhibiting recombinant SARS-CoV-2 helicase ATP/RNA-bound in its apoand conformations (46). The present study was carried out to investigate the potential application of inhibitors against SARS-CoV-2 NSP13 helicase.

MATERIALS AND METHODS

Sequence retrieval and homology modeling The amino acid sequence of SARS-CoV-2 NSP13 helicase (Accession number: YP_009725308) was retrieved from NCBI (601 aa) in FASTA format (22). PSI-BLAST was used for alignment against the Protein Databank (PDB) for a suitable template search with query sequence (23). The template with query sequence (PDB ID: 5WWP_A) was selected having a 72.20% identity score and 0.00 E-value. The SWISS-MODEL (https://swissmodel.expasy.org) structure assessment tool was used for homology modeling and visual analysis of the crystal structure of the protein (24).

PROCHECK server was used for validation of the structure model (25, 26), and its results suggested the reliability of the model (27). The overall G-factor and residue positions in φ - ψ plot regions analysis were used for the selection of a suitable model. The quality of the 3D model was verified by ERRAT, Verify3D, and ANOLEA. ERRAT plot confirmed the overall quality of the model (28).

Protein stability was analyzed by using SWISS-MODEL QMEAN (version 4.2.0) (29) and ProSA Z-score. The protein structure analysis (ProSA) calculated an overall quality score for the predicted structure (<u>30</u>).

Molecular docking

Molecular docking techniques dock small molecules into the protein binding site (<u>31</u>). The MCULE-1-Click docking (https://mcule.com) and InterEvDock 2.0 (https://bioserv.rpbs.univ-paris-diderot.fr/)

programs were used for docking calculations $(\underline{32})$. 1-Click docking is an online server for drug discovery platforms $(\underline{33})$.

The study protocol received approval from the Ethics Committee of the Pacific Institute of Medical Sciences, Sai Tirupati University, Udaipur, Rajasthan, India (Registration no: STU/2019/5006).

RESULTS

Protein model building

The alignment between the target and template was performed using PSI-BLAST (22). The 3D ribbon model of SARS-CoV-2 NSP13 helicase was generated using SWISS-MODEL (https://swissmodel.expasy.org) structure assessment tool (Figure 1).

Model reputation

The SARS-CoV-2 NSP13 helicase model corresponding to probability confirmation with 90.9% residue of the core section, 8.5% of the allowed section, and 0.4% residue of the outer section in the φ - ψ plot (34) (Figure 2a, b). The

above results indicate the reliability of protein models (<u>Table 1</u>) ($\underline{35}$, $\underline{36}$).

The verify-3D illustrates the compatibility of an atomic model (3D) with its amino acid sequence (1D) by assigning a structural class based on its location and environment (alpha, beta, loop, polar, and nonpolar) ($\underline{37}$). ANOLEA recorded non-local energy of the helicase was -6028 E/kT units, with a Z-score of -0.35. These scores indicate the good quality of the model (range 0.16 to -0.02) (Table 1).



Figure 1- The 3D ribbon structure model of NSP13 helicase

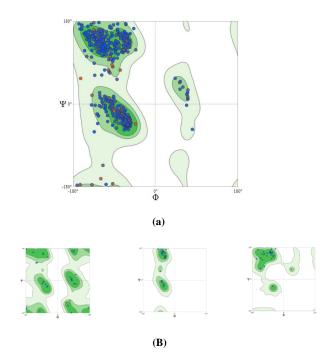


Figure 2-The φ-ψ plot of SARS-CoV-2 NSP13 helicase. (a) The total number of residues was 90.9% in the core, 8.5% in the allowed, and 0.4% in the generously allowed regions; (b) φ-ψ plot showing separate Φ and Ψ distributions of residues in glycine, proline, and pre-proline

Table 1- Evaluation of the protein model by PROCHECK VERIFY3D, and ANOLEA

Template		PROCHECK			Verify-3D	ANOLEA
NSP13	Core	Allowed	Generously outer	Disallowed	3D-ID Score	Z-Score
	90.9%	8.5%	0.4 %	0.2 %	95.9	-0.35

ERRAT analyses the statistics of non-bonded interactions between different atom types and plots the value of the error function versus position, which is calculated by comparison with statistics from highly refined structures ($\underline{38}$). ERRAT overall quality factor of the model was 93.5829, with an average probability value of 5.05729 (Figure 3).

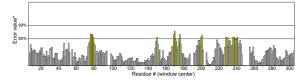


Figure 3- ERRAT result showing an overall quality factor of 93.5829 for the model (error-axis showing the error values to reject regions that exceed the error value)

Validation of the model

ProSA was used to determine the potential errors in the 3D model of SARS-CoV-2 NSP13 helicase. The archived ProSA Z-score of -9.17 indicates two aspects: overall model quality and energy deviation (Figure 4).

The QMEAN4 value of -0.99 was observed for the NSP13 helicase, which is very close to 0

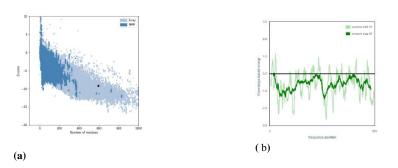


Figure 4- ProSA examination of SARS-CoV-2 NSP13 helicase overall model quality. (a) The blue dot in the plot shows the -9.17 z-score of predicted models; (b) The residue score plot shows energies of amino acids are less than zero, which represents good local model quality.

The QMEAN4 value of -0.99 was observed for the NSP13 helicase, which is very close to 0 and therefore an acceptable value (44). Assessed validity of the model predictable among 0 and 1, which could be concluded from the density plot locus set for QMEAN score (Figure 5). Figure 5 illustrates the QMEAN scores for the biological unit reference set, which were used as a tool for oligomeric protein assessment.

Molecular docking

The binding pockets of SARS-CoV-2 NSP13 helicase are not reported yet. Thus, the in silico approaches were utilized for the

prediction of binding pockets. MCULE-1docking (https://mcule.com) Click and InterEvDock-2.0 server were employed to explore the binding of ligands to the respective protein. The top five docking models of binding pockets of SARS-CoV-2 NSP13 helicase were identified and ranked based on the energy. More negative docking scores indicated higher binding affinity (Table 2). The summary table contains two rows: the ranks and docking energy scores from the input structures. Model 1 has high accuracy with an interface docking score of -5.940 kcal/mol from the crystal structure (Table 2).

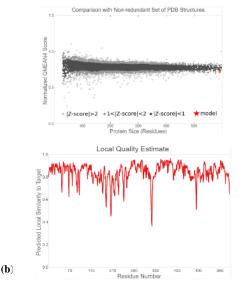


Figure 5- QMEAN scores for a biological unit reference set of SARS-CoV-2 NSP13 helicase. (a) Plot showing Z-score; (b) Local quality model for estimation of local summarily to target

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Molecular docking

The binding pockets of SARS-CoV-2 NSP13 helicase are not reported yet. Thus, the in silico approaches were utilized for the prediction of binding pockets. MCULE-1-Click docking (https://mcule.com) and InterEvDock-2.0 server were employed to explore the binding of ligands to the respective protein. The top five docking models of

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Table 2- Summary of the top five models. Row 1 (ranks of the models), Row 2 (docking energy scores - kcal/mol)

Rank	1	2	3	4	5
Docking Score	-5.940	-5.920	-5.790	-5.440	-5.430

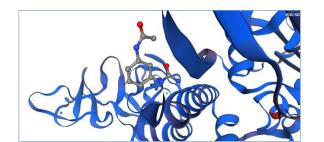


Figure 6- Binding pocket and interacting residues of the analyzed inhibitor (N-(3-(carbamoyl amino) phenyl) acetamide) using 1-Click docking server

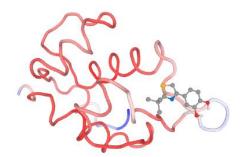


Figure 7- Binding pocket and interacting residues of the analyzed inhibitor (N-(3-(carbamoyl amino) phenyl) acetamide) using InterEvDock-2.0 server

Table 3- Top five binding pockets with energy and dimension

Ran	Energy		Center	
k	(Kcal/mol)	X-axis	Y-axis	Z-axis
1	-5.940	46.4820	13.1600	25.1190
2	-5.920	45.8500	14.4180	25.0310
3	-5.790	45.4730	15.1320	26.1920
4	-5.440	46.7390	12.5780	26.3880
5	-5.430	46.3440	13.2850	27.5460

The top five binding pockets of the SARS-CoV-2 NSP13 helicase enzyme were identified and ranked based on energy. The volume of the binding pockets was also analyzed in 3D dimension (Table 3).

The retrieved ligand N-(3-(carbamoylamino) phenyl) acetamide (PubChem CID: 828139) (Figure 7) for SARS-CoV-2 NSP13 helicase is

described in table 4.

The docking results revealed the Lys-146, Leu-147, Ile-151, Tyr-185, Lys-195, Tyr-224, Val-226, Leu-227, and Ser-229 residues exhibited good binding interactions with inhibitors and mutational studies of these residues could be highly effective in further studies.

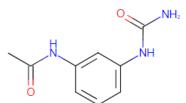


Figure 7- Structures of the selected inhibitor N-(3-(carbamoyl amino) phenyl) acetamide

Table 4- Properties and binding residues of the inhibitor N-(3-(carbamoyl amino) phenyl) acetamide (PubChem CID: 828139)

Ligand Properties	$C_9 H_{11} N_3 O_2$		
Molecular weight (g/mol)	193.203		
Component type	Non-polymer		
Hydrogen bond donor count	03		
Hydrogen bond donor count	02		
Rotatable bonds count	02		
Topological polar surface area	84.2 Å ²		
Heavy atom count	14		
Formal charge	0		
Interacting residues	Lys-146, Leu-147, Ile-151,		
-	Tyr-185, Lys-195, Tyr-224,		
	Val-226, Leu-227, Ser-229		

It was observed that the inhibitor binds at the binding residues between Lys-146 and Ser-229 (Table 4). The predicted structural and docking model described in this study may be further used for finding interactions with other SARS-CoV-2 enzymes to identify new anti-coronavirus targets.

DISCUSSION

In silico structural analysis revealed two ligand binding pockets on NSP13 that are the most conserved sites in the entire SARS-CoV-2 proteome (39). NSP13 ZBD conformations show the role of induced-fit flexibility in the ligand binding site (40). The SARS-CoV-2 NSP13 helicase model corresponds to probability confirmation with 90.9% residue of the core section that specifies the accuracy of the predicted model. In the verify-3D graph, 95.95% of the residues have averaged a 3D-1D score ≥ 0.2 which illustrates the results of good structures. ANOLEA recorded non-local energy of the helicase was -6028 E/kT units, with a Z-score of -0.35. These scores indicate the good quality of the model. ERRAT overall quality factor of the model was 93.5829 with an average probability value of 5.05729, which indicates a much more reliable and satisfactory model. The archived ProSA Z-score score of -9.17 also indicates two aspects: overall model quality and energy deviation. The values of the Z-score indicate fewer erroneous structures

(41, 42). Reliability of the projected model based on the scoring function of QMEAN that stated as 'Z-score' (43). Assessed validity of model predictable among 0 and 1, which could be concluded from the density plot locus set for QMEAN score. The QMEAN value comparison with the non-redundant protein collection revealed a different set of Z-values for different parameters. The diversion of the total energy of NSP13 helicase was measured by using Z-score (45). Drug hits have a significant role in inhibiting recombinant SARS-CoV-2 helicase in its apoand ATP/RNA-bound conformations (46). The potential compounds 26-deoxyactein and 25-O-anhydrocimigenol-3-O-beta-dxylopyranoside pose strong interactions and sustained close contact with NSP13 (47). The N-(3-(carbamoylamino) inhibitor phenvl) acetamide exhibited effective binding affinity against NSP13 helicase. The docking results revealed the Lys-146, Leu-147, Ile-151, Tyr-185, Lys-195, Tyr-224, Val-226, Leu-227, and

Ser-229 residues exhibit good binding interactions with inhibitor ligand N-(3-(carbamoyl amino) phenyl) acetamide. The predicted binding residues serve as the base for drug designing of lead compounds against SARS-CoV-2 NSP13 helicase, which is thought to play key roles during the replication of viral RNAs.

CONCLUSION

The results of this study establish N-(3-(carbamoylamino) phenyl) acetamide as a valuable lead molecule with great potential for SARS-CoV-2 NSP13 helicase inhibition.

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Ethics approvals and consent to participate

The study protocol received approval from the Ethics Committee of Pacific Institute of Medical Sciences, Sai Tirupati University, Udaipur, Rajasthan, India (Registration no: STU/2019/5006).

CONFLICT OF INTEREST

The authors declare that there is no conflict of interest regarding the publication of this article.

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