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In silico prediction of siRNA to silence the SARS-CoV-2 omicron variant targeting BA.4, BA.5, BQ.1, BQ1.1. and XBB: an alternative to traditional therapeutics

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Abstract

Background After the first infection in December 2019, the mutating strains of SARS-CoV2 have already affected a lot of healthy people around the world. But situations have not been as devastating as before the first pandemic of the omicron strains of SARS-CoV2. As of January 2023, five more Omicron offshoots, BA.4, BA.5, B.Q.1, B.Q.1.1 and XBB are now proliferating worldwide. Perhaps there are more variants already dormant that require only minor changes to resurrect. So, this study was conducted with a view to halting the infection afterwards. The spike protein found on the virus outer membrane is essential for viral attachment to host cells, thus making it an attractive target for vaccine, drug, or any other therapeutic development. Small interfering RNAs (siRNAs) are now being used as a potential treatment for various genetic conditions or as antiviral or antibacterial therapeutics. Thus, in this study, we looked at spike protein to see if any potential siRNAs could be discovered from it.

Results In this study, by approaching several computational assays (e.g., GC content, free energy of binding, free energy of folding, RNA–RNA binding, heat capacity, concentration plot, validation, and finally molecular docking analysis), we concluded that two siRNAs could be effective to silence the spike protein of the omicron variant. So, these siRNAs could be a potential target for therapeutic development against the SARS-CoV2 virus by silencing the spike protein of this virus.

Conclusion We believe our research lays the groundwork for the development of effective therapies at the genome level and might be used to develop chemically produced siRNA molecules as an antiviral drug against SARS-CoV2 virus infection.

Keywords SARS-CoV2, siRNA, Spike protein, Small interfering RNA, Molecular docking, In silico analysis

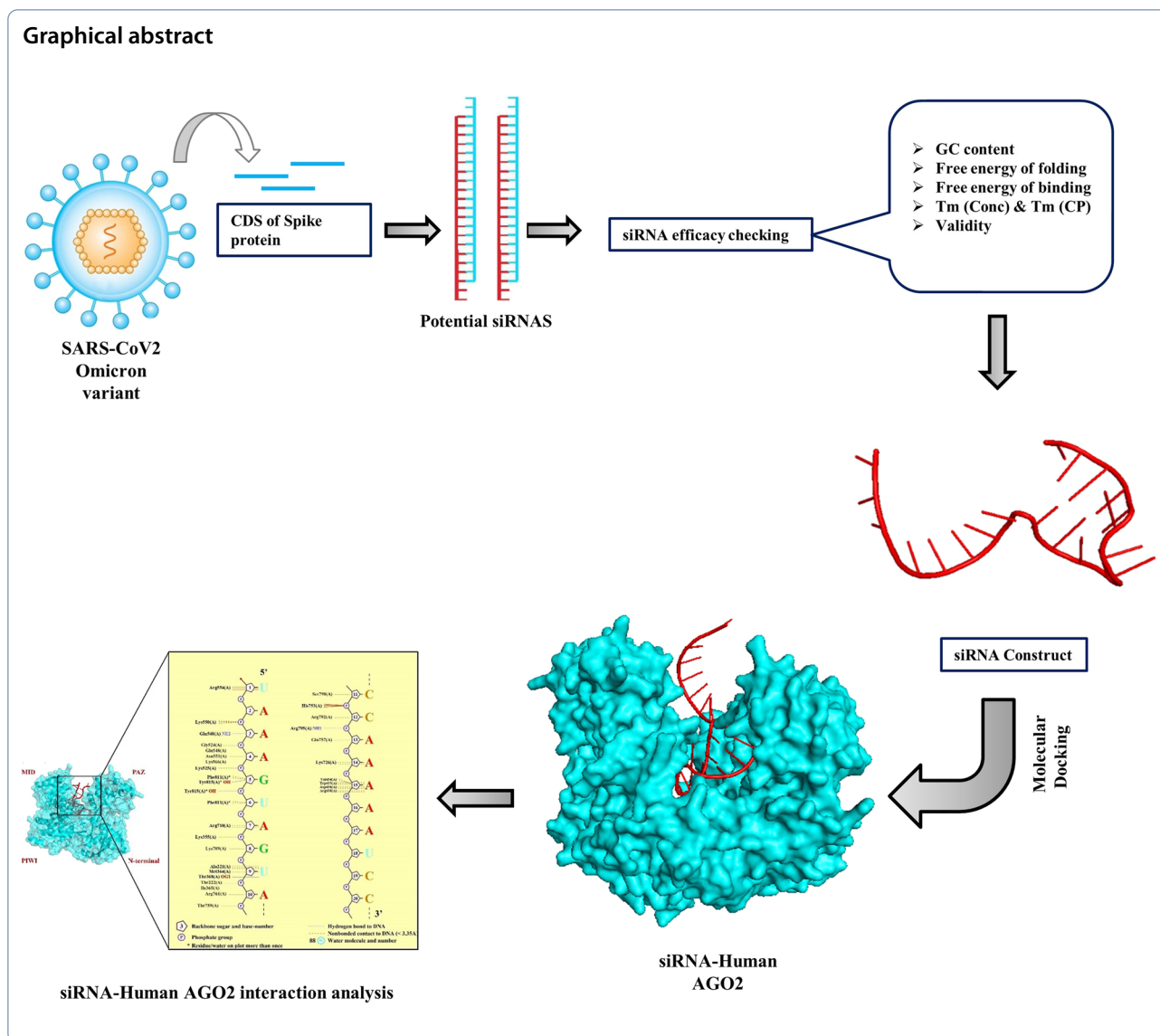
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Background

Severe acute respiratory syndrome coronavirus-2 (SARS-CoV-2), a new Beta coronavirus, was discovered in December 2019, infecting more than 0.318 billion people worldwide and resulting in 5.5 million deaths [1, 2]. Different strains (Alpha, Beta, Gamma, Delta, Kappa, and so on) of this virus have developed during the pandemic due to mutations in the SARS-CoV-2 genome. Both the antigenic evasion mechanism as well as the rate of infection are more affected by these modifications [3, 4]. The advent of numerous variations causes waves of destructive pandemics to spread worldwide [5]. The World Health Organization (WHO) designated this new type as a variation of concern on November 26, 2021, after it was discovered on November 24, 2021, in Botswana, South

Africa (variant of concern- VOC) [6]. The appearance of this highly modified SARS-CoV-2 strain (B.1.1.529, Omicron) and its fast spread across six continents within a week of its discovery have heightened global public health concerns [7]. After that, this variant changed several times, causing several spikes worldwide. As of June 2022, two more Omicron offshoots, BA.4 and BA.5 were started to proliferate worldwide [8]. Also, when compared to the BA.1 strain, which sparked the Omicron wave in most countries late last year, these two variants of concern (VOC) are far more similar to BA.2. However, from December, 2022 these two subvariants were started to replace by newer omicron subvariants, BQ.1, BQ.1.1 and XBB specifically in United States of America. Now, as of January 2023, the variants that are proliferating worldwides are BA.4, BA.5, BQ.1, BQ.1.1 and

XBB.1.5 with the most potent distribution of BQ.1 pango lineage which is 49.0% [9, 10]. The major mutations that are found in each of the sublineages are follows: S135R (NSP1), F486V (S-protein), L11F (ORF7b) and P151S (N-protein) for BA.4. For BA.5 the major mutations are S135R (NSP1) and F486V (S-protein). For BQ.1 defining mutations are: Y272H (RdRP), M233I (NSP13), K444T (S-protein) and N460K (S-protein). For BQ.1.1 the mutations are similar to the BQ.1 except to two new mutations which are N268S (NSP13) and R346T (S-protein). The all defining mutations in XBB.1.5 are from spike proteins which are V83A (S-protein), H146Q (S-protein), Q183E (S-protein), G252V (S-protein), F486L (S-protein) and F486S (S-protein) [11].

Alphacoronavirus, Betacoronavirus, Gammacoronavirus, and Deltacoronavirus are the four genetically distinct groups of coronaviruses [12]. Mammals are the primary hosts for the first two species, whereas birds are the hosts for the latter two. Coronavirus genomes are generally 26–32 kb in size and include 6–11 open reading frames (ORFs) [13]. The four essential structural proteins of coronavirus are nucleocapsid protein (N), a small envelope protein (E), spike surface glycoprotein (S), and matrix protein (M), and each one is necessary to produce a physically complete virus [14, 15].

The human angiotensin converting enzyme 2 (hACE2) present in lung cells is recognized and can bind to the Spike glycoprotein (S), which starts the pathogenesis of SARS-CoV-2 when viral particles adhere to host cell cellular surface receptors [2, 16]. So, virus entry into cells is made possible by this spike glycoprotein (S), which forms homotrimers on the viral surface [17, 18]. S consists of two functioning parts, S1, and S2 subunits. The membrane-anchored S2 subunit, which contains the fusion machinery, is stabilized by the S1 subunit, which contains receptor-binding domains [19]. Because coronavirus S glycoprotein is surface-exposed and aids host cell entry, it may be recognized as a therapeutic target [20]. Previous research, however, found that all identified variations had most of the mutations in their Spike glycoprotein [21]. As a result, finding an antiviral medication that targets Spike protein is a promising option for blocking COVID-19 variant transmission [22].

A biological regulatory process called RNA interference (RNAi) uses post-transcriptional gene silencing to silence mRNA. RNAi is a technology that also shows promise for reducing human viral infections [23, 24]. Non coding RNA like small interfering RNAs (siRNAs) can inhibit gene expression by hybridizing to complementary mRNA and neutralizing it [25]. The siRNA is a 19–25 base pair long RNA duplex with two nucleotide overhangs on the 3' end. It binds to target complementary mRNA and

degrades its enzymatic quality to trigger post-translational gene silencing (PTGS) [26].

However, the process of binding of the siRNA's with complementary mRNA is not an easy task. The siRNA-mediated inhibition of gene expression is a very intricate process. After entering the cell, the siRNA duplex is splitted by dicer, an enzyme similar to RNase III, and integrated into the protein complex known as the RNA-induced silencing complex (RISC) [27, 28]. The RNA strands within RISC are divided by the ATP-dependent RNA helicase domain. RISC eliminates the target mRNA's sense strand, but the catalytic RISC protein, an argonaute protein, can align RISC on the target mRNA and cleave the target mRNA's strand [29].

Following the discovery of its mechanism, this method has evolved into a robust experimental gene-silencing tool in fundamental research [30]. For instance, studies show that combining chemically made siRNA duplexes that target SARS-CoV genomic RNA leads to up to 80% virus suppression [31]. Because the spike protein produced by the S gene of SARS-CoV2 omicron variant is part of the cell surface entry complex, it can be used as a viable target for suppressing SARS-CoV-2-induced infection. In this paper, siRNA molecules for the SARS-CoV-2 "S" gene were rationally generated using various computational methods. Designed siRNAs might aid in the discovery of effective treatment medicines against this killer virus. The RNAi therapies GIVLAARITM and ONPATPRO® are now recognized for use in treating acute hepatic porphyria and polyneuropathy, respectively [32]. We believe this research will aid in developing a similar therapeutic technique for SARS-CoV-2.

Methods

The complete methodology for designing of the siRNA molecules against the SARS-Cov2 omicron variant is shown in the graphical abstract.

Sequence retrieval of CDS of spike genome

The National Health Laboratory South Africa reported the Omicron spike genomic sequence to the Global Initiative for Sharing all Influenza Data (GISAID) with the accession number EPI_ISL_8616776. We selected this variant as this is the first reported sub lineage of omicron variant. Then we identified and sorted out the Spike CDS (coding sequence) from the EPI_ISL_8616776 through NCBI BLAST search.

Designing of siRNA from the CDS of spike genome

To identify the siRNA molecule from the CDS of spike genome, siDirect version 2.0 webserver was used (<https://sidirect2.rnai.jp/doc/>) [33]. To achieve this, first of all the retrieved fasta sequence of omicron variant spike genome

submitted in the siDirect web server. The web server then identified potential siRNAs employing three rules: Ui-Tei, Renold, and Amarguioui [34–36]. The seed duplex's melting temperature (T_m) is by default, set lower than 21.5 °C on the webserver. It is essential because the seed duplex melting temperature influences the efficacy of siRNAs, such as off-target effect reduction [33]. The equation to calculate melting temperature is below:

$$T_m = (1000 \cdot \Delta H) / (A + \Delta S + R \ln(CT/4)) - 273.15 + 16.6 \log [Na^+].$$

Here,

- The sum of the nearest neighbor enthalpy change is represented by ΔH (kcal/mol)
- The helix initiation constant (−10.8) is represented by A
- ΔS represents the sum of the nearest neighbor entropy change
- The gas constant (1.987 cal/deg./mol) is characterized by R
- The total molecular concentration (100 μ M) of the strand is represented by CT and
- The concentration of Sodium, $[Na^+]$ was fixed at 100 mM

The three algorithms chosen for siRNA prediction each have unique characteristics. For example, the Ui-Tei algorithm follows specific rules, such as (i) 5' terminus of the antisense/guide strand has to include A/U nucleotide, (ii) 5' end of the sense/passenger strand must contain G/C nucleotide, (iii) 5' terminal 7 base pairs of sense/passenger strand has to contain at least 4 A/U nucleotides, and (iv) GC stretch should not be longer than nine nucleotides [34]. Meanwhile, Amarguioui rules include the parameters such as (i) robust binding of 5' sense/passenger strand, (ii) the A/U differential of the duplex end should be more than zero, (iii) position six should always contain A, (iv) position one must contain any base except U, (v) weak binding of 3' sense/passenger strand and (vi) position 19 must contain any base except G [35]. Reynolds algorithm also follows several criteria, such as (i) the sense/passenger strand must maintain ≥ 3 base pairs at the position between 15 and 19, (ii) maintenance of GC content in the designed siRNA between 30 to 52%, (iii) position 19 and 3 of the sense/passenger strand must contain A, (iv) internal stability has to be low at a target site, (v) sense/passenger strand should contain U at position 10, (vi) position 13 of the sense/passenger strands must contain any bases other than G [36].

Investigation of parameters for siRNA refinement

To identify the most effective siRNA's from the bulk siRNAs that was initially reported through siDirect

webserver, we used several refinement procedures for highly effective siRNA selection. First, GC content of the siRNA molecules was calculated through the OligoCalc web server (<http://biotools.nubic.northwestern.edu/OligoCalc.html>) [37]. Any siRNA's that showed GC content under 30% were excluded from the study. Next, the secondary structure and free energy of folding of the siRNAs were predicted using the RNA structure website <https://rna.urmc.rochester.edu/RNAstructureWeb/> [38]. We excluded from further analysis of any siRNAs that displayed negative free energy of folding in the website. Then, we anticipate the interaction of the target and guide strands of siRNAs with RNA. The thermodynamic interaction between the target and guide strands was consequently calculated using the Bifold tool of the RNA structure website [33]. The heat capacity and concentration charts were then created using the DINA Melt web server (<http://www.unafold.org/hybrid2.php>) [39]. The melting temperature T_m (Cp) is displayed in the detailed heat capacity figure along with the ensemble heat capacity (Cp) as a function of temperature. The melting temperature T_m (Conc), which may be determined using the concentration plot, is obtained at the point where double-stranded molecules' concentration is half their maximum value. Finally, SMEpred webserver (<https://bioinfo.imtech.res.in/manojk/smepred/>) was used to validate the final siRNAs [40]. SMEpred is the world's first website for designing and predicting the efficiency of chemically modified siRNAs. The anticipated siRNAs are tested on different datasets: standard siRNAs dataset (2182) and cm-siRNA dataset (3031 cm-siRNAs), both of which have been experimentally validated. SMEpred was also used to do a tenfold cross-validation employing Support Vector Machine (SVM).

Conservancy checking against the other sub-lineages and human genomic transcript

In the final step of siRNA prediction, a conservancy checking was performed against the 59 sub-lineages of the omicron variant through NCBI Blastn search [41] and multiple sequence alignment through CLC Drug Discovery Workbench 3.0 software. In the NCBI Blastn database, we manually uploaded the spike CDS of all sub-lineages and all other parameters were selected as default for Blast search. For phylogenetic tree construction, we employed a neighbor-joining phylogenetic tree with a bootstrap value of 1000. First, the phylogenetic tree was generated using the Tamura Nei assessment model [37]. Then, the phylogenetic tree was constructed using the MEGA-11 tool [38]. Finally, we used NCBI Blastn to perform a specific blast analysis to compare the generated siRNAs to human genomic transcripts. The e-value was adjusted to $1e-10$ to lessen the search criterion's

stringency and hence increase the likelihood of arbitrary matches.

Molecular docking of guide siRNA and argonaute-2 protein

The right interaction between siRNA duplex (primarily guide strand) and RISC complex protein (mostly human argonaute protein) is required to initiate an adequate antiviral response via RNAi-mediated viral gene silencing [30]. After eliminating the target mRNA's sense strand, the catalytic RISC protein, which is the argonaute protein, can align RISC on the target mRNA and cleave the target mRNA's strand [29]. That is why docking of the siRNA with argonaute protein is an indicator of successful RISC complex formation and siRNA efficacy. So, we docked our siRNA's with argonaute-2, one of the protein of RISC complex.

Molecular docking of the siRNA guide strand with argonaute-2 protein was conducted with HDOCK web server [42]. Before molecular docking we predicted the 3D model of the siRNAs and argonaute-2 protein. For identifying the 3D structure of human argonaute-2 Robetta webserver was used [43]. This homology modeling webserver employs deep learning algorithms, RoseTTAFold and TrRosetta, and an integrated reporting facility for specific sequence alignments for homology modeling. For predicting the 3D structure of siRNA guide strand, we used Mfold and RNA Composer webserver [44, 45]. For predicting the 3D structure of siRNA guide strand, we used Mfold and RNA Composer web server [44]. The mfold web server, used to calculate the folding pattern of DNA/RNA at 37 °C, is one of the oldest known online servers in computational molecular biology. The RNA Composer system, on the other hand, provides a new user-friendly technique to fully autonomous modeling of immense RNA 3D structures. The method relies on the automatic translation concept and uses the RNA FRABASE database as a lexicon to connect RNA secondary and tertiary design components. Finally, after modeling of the guide siRNA and human argonaute-2 protein, we docked the siRNA with RISC complex (argonaute-2) through molecular docking. After docking, we visualize the interaction pattern through the PDBsum web server [46]. Web server PDBsum provides structural data on Protein Data Bank entries (PDB), protein secondary structure, protein–ligand, and protein–DNA.

Results

Sequence retrieval and 702 siRNA prediction through SiDirect

The complete CDS of omicron spike protein was retrieved from the EPI_ISL_8616776 by blast searching against Sars-cov2 genomic data. After that, the siDirect

webserver was used to identify the potential siRNA's from the CDS of the omicron spike protein. siDirect used several parameters, including Ui-Tei, Renold and Amarguioui rules to identify potential siRNA's with melting temperatures below 21.5 °C to reduce the seed-dependent off-target binding. Initially, siDirect webserver predicted 702 potential siRNA's from CDS of the omicron spike protein. We then filtered this 702 siRNA's to 17 siRNA's by combining the three parameters (Ui-Tei, Renold and Amarguioui rules) and by selecting those siRNA's whose melting temperature is below 10 °C. So, this 17 siRNA's are highly off target reduced siRNA's (Table 1).

GC content Calculation of the predicted 17 siRNA's

The amount of GC content in the indicated 17 siRNA molecules was identified through the GC-content calculator (Additional file 1: Table S1). However, we found only 5 siRNAs showing GC content greater than 30% after prediction. Therefore, to be a potential siRNA, the GC content of the siRNA's must be ranged from 30 to 60% [47]. We then filtered the rest 12 siRNAs from this study as the GC content of those siRNA's were less than 30%.

Secondary structure prediction of the 5 siRNA's

The calculated free energy of folding as well as the secondary structure of the 5 siRNA's was predicted through RNA Structure webserver. The calculated free energy of folding of the 5 siRNA's ranged from –1.4 to 1.8 (Additional file 1: Table S2 and Fig. 1). Among them siRNA target number S3 showed no binding pairs for secondary structure prediction. However, only two targets, e.g., S2 and S10 showed positive free energy of folding after calculation. Those two siRNA's were selected for further studies as these siRNAs are counted as less prone to folding.

Computation of RNA–RNA binding, heat capacity, concentration plot, and validation

The free energy of hybridization between the guide and target strand of the final two siRNA's was computed. For S2 and S10 the free energy of binding was calculated as –30.2 and –28.4, respectively (Fig. 2). After, we calculated the heat capacity (TmCp) and duplex concentration (TmConc). The more these melting temperature values the better is the candidate molecules. The Tm(Cp) and Tm(Conc) of S2 molecule were calculated as 81.4 and 80.2, respectively, which is slightly more significant than the Tm(Cp) and Tm(Conc) of S10, e.g., 76.3 and 75.1, respectively (Additional file 2: Fig. S1). Finally, we validated both of the siRNA molecules by checking the effectivity through SMEpred webserver. The web server calculated better candidacy for S2 molecule with a score

Table 1 siRNA's with target sequences predicted by siDirect webserver. Here all these siRNA's follows the all three rules of siRNA selection, e.g., Ui-Tei, Reynolds and Amarzguioui

| Alias | Target Sequence (21nt target + 2nt overhang) | Target position in mRNA of Spike | RNA oligo sequences 21nt guide (5' → 3') 21nt passenger (5' → 3') | Functional siRNA selection: Ui-Tei Reynolds Amarzguioui | Seed duplex stability (Tm) | |
|-------|---|----------------------------------|---|--|----------------------------|-----------|
| | | | | | Guide | Passenger |
| S1 | ATGTTTGTTCCTGTTTTATT | 1–23 | UAAAACAAGAAAAACAACAU GUUUUUUUUUUUUUUUUUUU | U R A | 5.6 °C | 5.6 °C |
| S2 | TGGATTTTGGTACTACTTTAGA | 304–326 | UAAAGUAGUACCAAAAAUCCA GAUUUUUGGUACUACUUUAGA | U R A | 9.8 °C | –3.3 °C |
| S3 | CGCTACTAATGTTGTTAAAG | 360–382 | UUAUAACAACAUUAGUAGCG CUACUAAUGUUUUUUUUAAAG | U R A | 1.4 °C | 6.3 °C |
| S4 | GGAATTTGTGTTAAGAATATT | 554–576 | UAUUUUAAAACACAAAUUCCC GAAUUUGUUUUUAGAAUUAU | U R A | 6.9 °C | 5.3 °C |
| S5 | AAGAATATTGATGTTATTTTAA | 568–590 | AAAAUAACCAUCAAUUUCUU GAAUUAUGAUGGUUUUUUUAA | U R A | –0.3 °C | –1.8 °C |
| S6 | GACTTTTCTATTAATAATAATG | 810–832 | UUUAUUUUUAUAGAAAAGUC CUUUUCUUAUUAAAUAUAUUG | U R A | –8.0 °C | 7.1 °C |
| S7 | TTCTATTAATAATAATGAAAAT | 815–837 | UUUCAUUUAUUUUUAUAGAA CUAUUUAAAUAUAUAGAAAUAU | U R A | 8.9 °C | –7.5 °C |
| S8 | ATCTATCAAACCTCTAACCTTAG | 925–947 | AAAGUUAGAAGUUUGAUAGAU CUAUCAAACUUCUACUUUUG | U R A | 9.8 °C | 8.9 °C |
| S9 | TTGTTAGATTCCTAATATTACA | 968–990 | UAAUUAUAGGAAAUUCAACA GUUAGAUUCCUAAUUAUACA | U R A | –8.0 °C | 6.9 °C |
| S10 | CCCTTTTGATGAAGTTTTTAACG | 999–1021 | UUAAAAUCUCAUCAAAGGG CUUUUGAUGAAGUUUUUUAACG | U R A | 0.0 °C | 7.4 °C |
| S11 | CTGGAAATATTGCTGATTATAAT | 1235–1257 | UAUAUCAGCAAUUUUCCAG GGAAAUUUGCUGAUUUAUUAU | U R A | 8.7 °C | 1.8 °C |
| S12 | GACCTAAAAAGCTACTAATTTG | 1568–1590 | AAUUAGUAGACUUUUUAGGUC CCUAAAAAGUCUACUAAUUUG | U R A | 6.3 °C | –3.8 °C |
| S13 | GTCTACTAATTTGTTAAAAACA | 1578–1600 | UUUUUAACCAAUUAUAGUAGAC CUACUAAUUUGGUUAAAAACA | U R A | 0.0 °C | 6.3 °C |
| S14 | TTGCAATATGGCAGTTTTTGTAC | 2251–2273 | ACAAAAACUGCCAUAUUGCAA GCAAUAUGGCAGUUUUUGUAC | U R A | 5.6 °C | 5.6 °C |
| S15 | CAGTTTTTGTACACAATAAAAC | 2262–2284 | UUUAUUUGUACAAAAACUG GUUUUUGUACACAUAUAAAAC | U R A | –1.4 °C | 5.6 °C |
| S16 | GTGCAATTTCAAGTGTTTTAAAT | 2903–2925 | UUAAAACACUUGAAAUUGCAC GCAAUUUCAAGUGUUUUAAAUAU | U R A | 7.2 °C | 7.4 °C |
| S17 | TTGTAACATTCAAAAAGAAATT | 3518–354 | UUUCUUUUUGAAUGUUUACAA GUAACAUAUCAAAGAAAUU | U R A | 5.5 °C | 6.9 °C |

of 89.5 rather than S10 molecule, which was calculated as 78.2 (Table 2). This result resolved that both siRNA molecules could be effective for advanced molecular docking.

Calculation of off-target effect and conservancy search against other sub-lineages

Standalone blast search for both siRNA molecules was conducted against human transcriptome genome to find out possible homology. This results in our predicted siRNA's being unique and not interacting with any off-target genome other than viral targets. After that, we employed a conservancy analysis of these 2 siRNA molecules against 59 sub-lineages of omicron variants

through multiple sequence alignment NCBI Blast search. S2 molecule showed 100% conservancy, whereas S10 molecule showed 96% conservancy (out of 59, 57 sub-lineages matched with S10) (Additional file 3: Fig. 2). We also build a phylogenetic tree of the 59 sub-lineages for Spike protein (Additional file 4: Fig. S3). Only a handful number of lineages Showed significant divergence after tree analysis (bootstrap value >0.74). These results stated that our predicted siRNA's targets are mostly conserved among the other sub-lineages of omicron variant (Table 2).

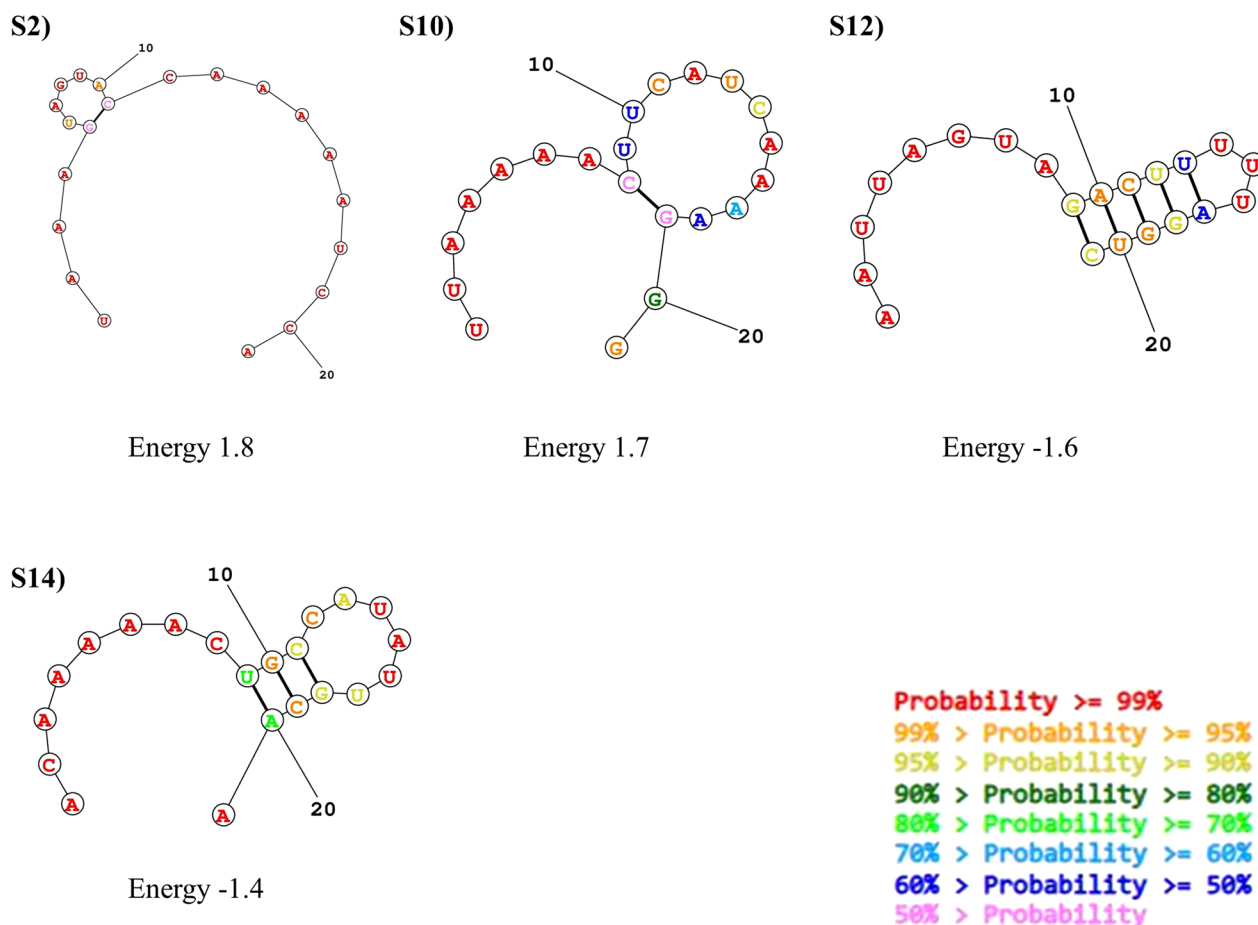


Fig. 1 Prediction of free energy of folding of the putative siRNAs. (S2), (S10), (S12) and (S14) consecutively denoted the guide strand of siRNA molecules S2, S10, S12 and S14. Among them only siRNA S10 and S10 showed positive free energy calculated by RNA Structure webserver

Molecular modeling and docking analysis of final siRNA's & Ago2

Mfold and RNA Composer webserver conducted molecular modeling of the final siRNA molecules. First, we attributed the guide siRNA sequence in the Mfold webserver to form the RNAdraw format. Then we used this RNAdraw format in the RNA Composer webserver to compose the final 3D structure of the final siRNA molecules. After designing the 3D models of siRNAs, we modeled the 3D structure of the human Ago2 (argonaute 2) protein through Robetta homology modeling web server. We used the refseq sequence of the human Arg2, e.g., UniprotKB: Q9UKV8 to model the Ago2 protein. The template for the homology modeling was selected was 4Z4D crystal structure (Human argonaute protein bound to t1-G target RNA) as this protein showed maximum sequence similarity with our Ago2 sequence. The modeled protein was then refined in the GalaxyRefine webserver. Finally, the quality of the model was checked using Ramachandran plot analysis of ZLab webserver [48].

Ramachandran's analysis of Ago2 3D structure revealed a good plot with 99.062% residues in the highly preferred observation. Only 0.938% of residues were found in the preferred region and no residues were found in the questionable region (Fig. 3).

Finally, molecular docking of the final siRNA molecules (S2 and S10) and human Ago2 was conducted by HDock webserver. We then selected the best-docked complex with low energy for interaction analysis through Pymol and PDBsum webserver. Docking analysis revealed slightly better interaction of S10 molecule with Ago2 (docking score -350.23); for S2 molecule the docking score was found as -300.17 (Table 3). These siRNA's are docked in the same pocket spanning between the Paz, Mid and PIWI domains of Ago2. The PDBsum analysis of both docked complexes S2 and S10 is shown in Figs. 4 and 5 respectively.

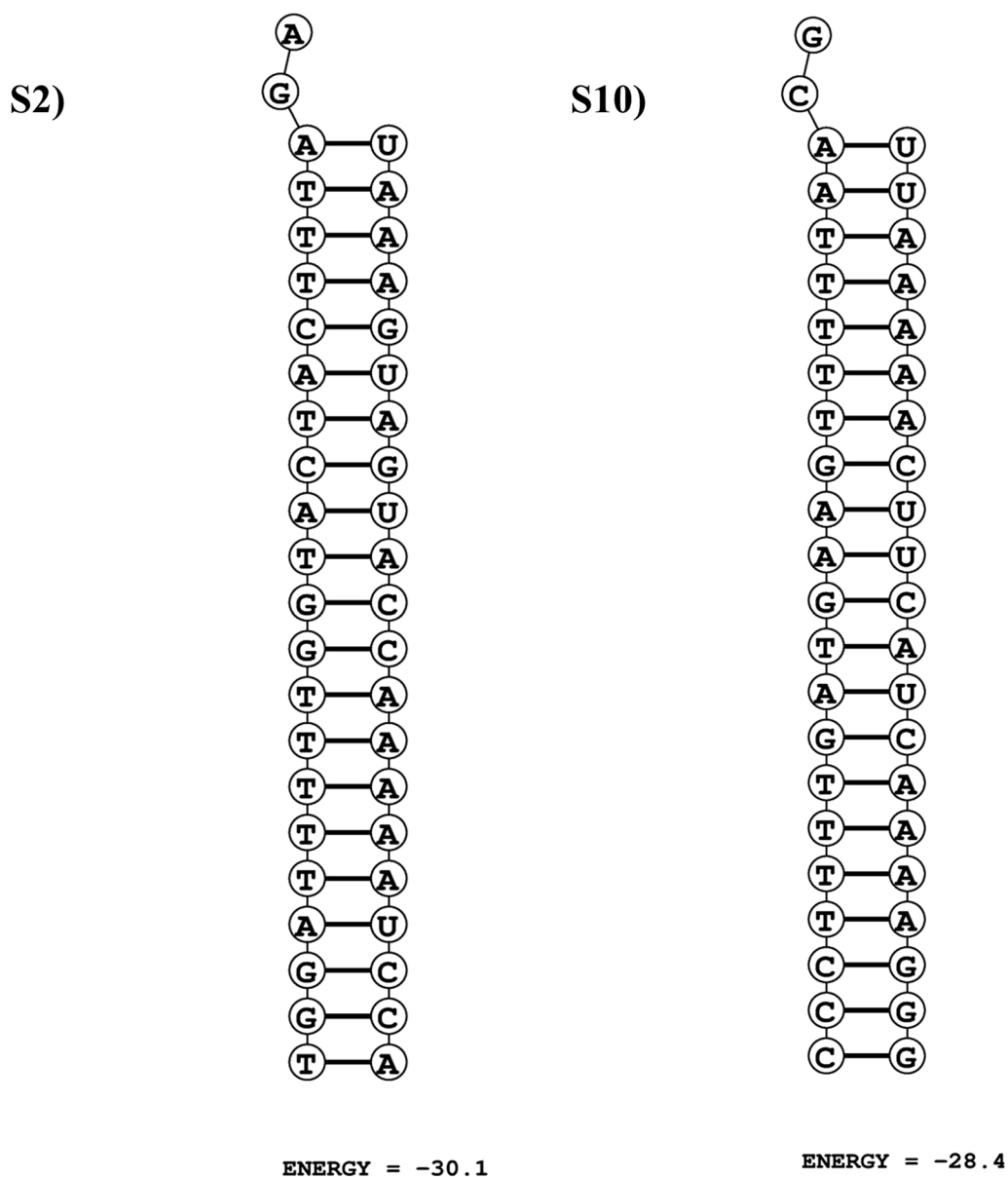


Fig. 2 Prediction of free energy of binding of the putative siRNA's with target RNA. (S2) denoted the siRNA molecule S2 and (S10) denoted the siRNA molecule S10. Both of this siRNA molecule (guide strand) showed greater binding efficacy with the target RNA strand

Discussion

COVID-19 was considered a pandemic, and mortality rates increased as a result of multiple transmissible variants (e.g., delta/B.1.1.7.2; alpha/ B.1.1.7; gamma/P.1; beta/B.1.351 and as of January 2023, now Omicron sub variant B.4, BA.5, BQ.1, BQ.1.1 and XBB.1.5). BA.4 and BA.5 now account for more than 21% of new cases in the U.S., according to U.S centers for Disease Control and Prevention (CDC) estimates [49]. Also, according to the ECDC (European Centre for Disease Prevention and Control), as of 8 January 2023, the estimated distribution

of variants of concern (VOC) or variants of interest (VOI) was 49.0% (from eight countries) for BQ.1, 24.3% (from 10 countries) for BA.5, 12.7% (from 10 countries) for BA.2.75, 2.7% (from eight countries) for XBB, 1.4% (from six countries) for XBB.1.5, 0.8% (from eight countries) for BA.2, and 0.8% (from nine countries) for BA.4 [9]. While, on the other hand, in USA the dominant variant nationwide is the XBB.1.5 with 43% cases. The second dominant sublineage in USA is the BQ.1.1 with 29% cases. According to researchers, these novel subvariants developed from the Omicron lineages to become much

Table 2 Calculation of free energy of binding, heat capacity, concentration plot and conservancy search against other sub-lineages of Sars-CoV2 virus

| Alias | Target position in mRNA of Spike | Target Sequence (21nt target + 2nt overhang) | RNA oligo sequences 21nt guide (5' → 3') 21nt passenger (5' → 3') | Free energy of binding | Tm | | Validity | % of RNA sequence matched with sub-lineages |
|-------|----------------------------------|--|---|------------------------|--------|-----------|----------|---|
| | | | | | Tm(Cp) | Tm (Conc) | | |
| S2 | 304–326 | TGGATTTTGGTACTACT TTAGA | UAAAGUAGUACCAAA AAUCCA GAUUUUUGGUACUAC UUUAGA | −30.2 | 81.4 | 80.2 | 89.5 | 100% (59/59) |
| S10 | 999–1021 | CCCTTTTGATGAAGTTTT TAACG | UUAAAAACUUCAUCA AAAGGG CUUUUGAUGAAGUUU UUAACG | −28.4 | 76.3 | 75.1 | 78.2 | 96.61% (57/59) |

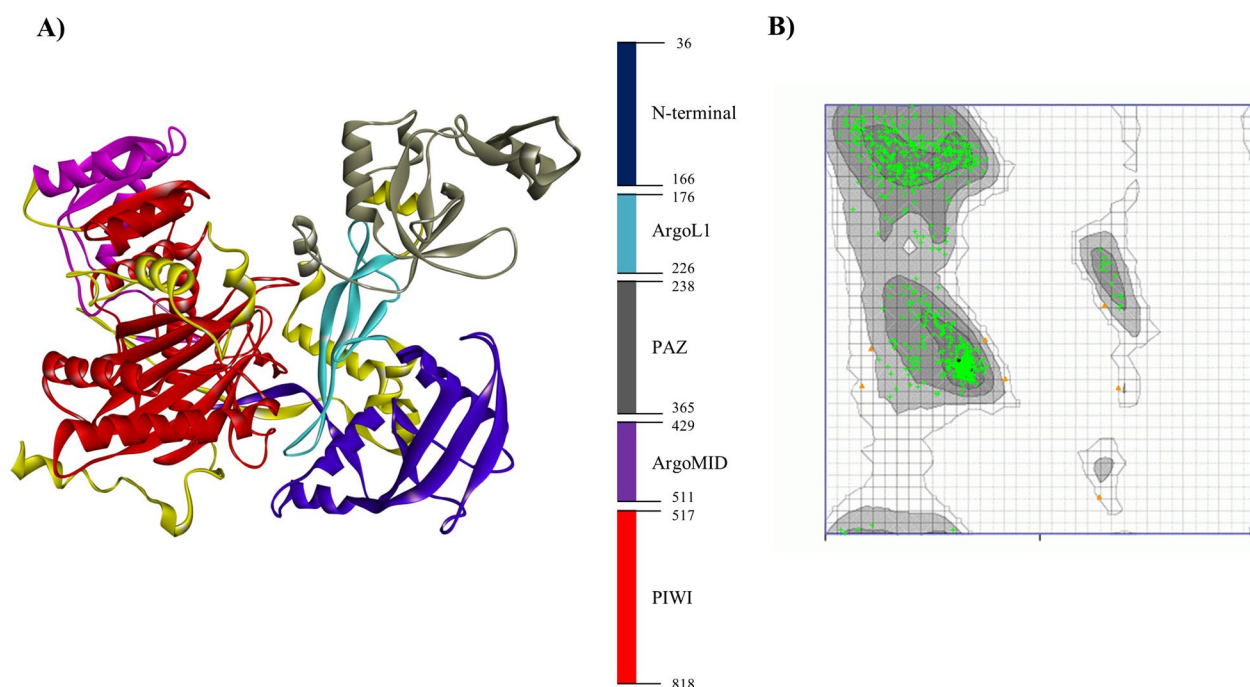


Fig. 3 (A) Homology modeling and (B) Ramachandran plot analysis of human argonaute-2 protein. Ramachndran plot analysis revealed 99.062% residues in the highly preferred observation, 0.938% residues in the preferred region and no residues in the questionable part

more contagious and can circumvent immunity from a prior infection [8]. Vaccine-induced antibodies are consistently more successful at blocking earlier Omicron strains, such as BA.1 and BA.2, than they are at blocking BA.4 and BA.5 [50–52]. So, a newer treatment method is now at the peak of concern in the scientific research community.

The Omicron variant, B.1.1.529 was first detected in South Africa’s Gauteng area in November 2021 [53]. This Variant of Concern (VOC) has different epidemiological changes in transmission rate and at least 32 mutations in its spike protein, which may be affecting the current pandemic trajectory [54]. Without regard for region,

this pandemic has a global reach; for example, this type has infected nearly 60 countries, and no continent is protected [55]. Furthermore, there is no effective vaccine to prevent the omicron variant, and no RNAi-based treatment is now in use or has been developed. A successful vaccination campaign in early 2021 significantly increased population immunity; however, the emergence of the delta or omicron lineages of SARS-CoV-2 has posed a new challenge to immunization-delaying methods [56, 57]. The current vaccinations are based on the SARS-CoV-2 Wuhan strain; however, the virus no longer looks like that [55]. Researchers are currently relying mainly on omicron sequencing data, which reveals a

Table 3 Docking interaction analysis of the best binding complex S2 and S8 siRNA with human Argonaute-2

| Alias | RNA oligo sequences 21nt guide (5' → 3') | Docking score | Interacting residues in Ago2 domains | | | | |
|-------|---|---------------|--------------------------------------|---|---|--------------------------------------|---|
| | | | N-terminal (36–166) | L1 (176–226) | PAZ (238–365) | Mid (429–511) | PIWI (517–818) |
| S2 | UAAAGUAGUACCAAAAUCCA | −300.17 | N/A | ALA221*# Thr222*# | LYS355 THR361 MET364 ILE365*# THR368 ARG375 | VAL434 TRP435 ASP436 ARG438 | GLY524*, LYS525 GLN548, LYS550 ASN551*, GLN553 ARG554, GLN558 ASN562, LYS566* THR599, HIS600 PRO601, PRO602 ALA603, LYS709*# ARG710*#, LYS726 HIS753, ALA754 ILE756, GLN757*# THR759*#, SER760 ARG761*#, ARG792*# ARG795, SER798*# HIS807, PHE811 TYR815 |
| S10 | UUAAAAUCUCAAAAAGGG | −350.23 | N/A | ASP218 VAL219 SER220* ALA221*# THR222*# ALA223 | ARG351* CYS352 LYS355 LEU356 THR357 ASP358 GLN360 THR361 MET364 ILE365*# THR368 ALA369 ARG375 | VAL434 TRP435 ASP436 ARG438 | LYS550, GLN558 ASN562, LEU565, LYS709*#, ARG710*# ARG714, LYS720 GLY725, LYS726 HIS753, ALA754 GLY755, ILE756 GLN757*#, GLY758 THR759*#, SER760 ARG761*#, TYR792*# CYS793, THR794 ARG795, SER796 VAL797, SER798*# ILE799, TYR804* |

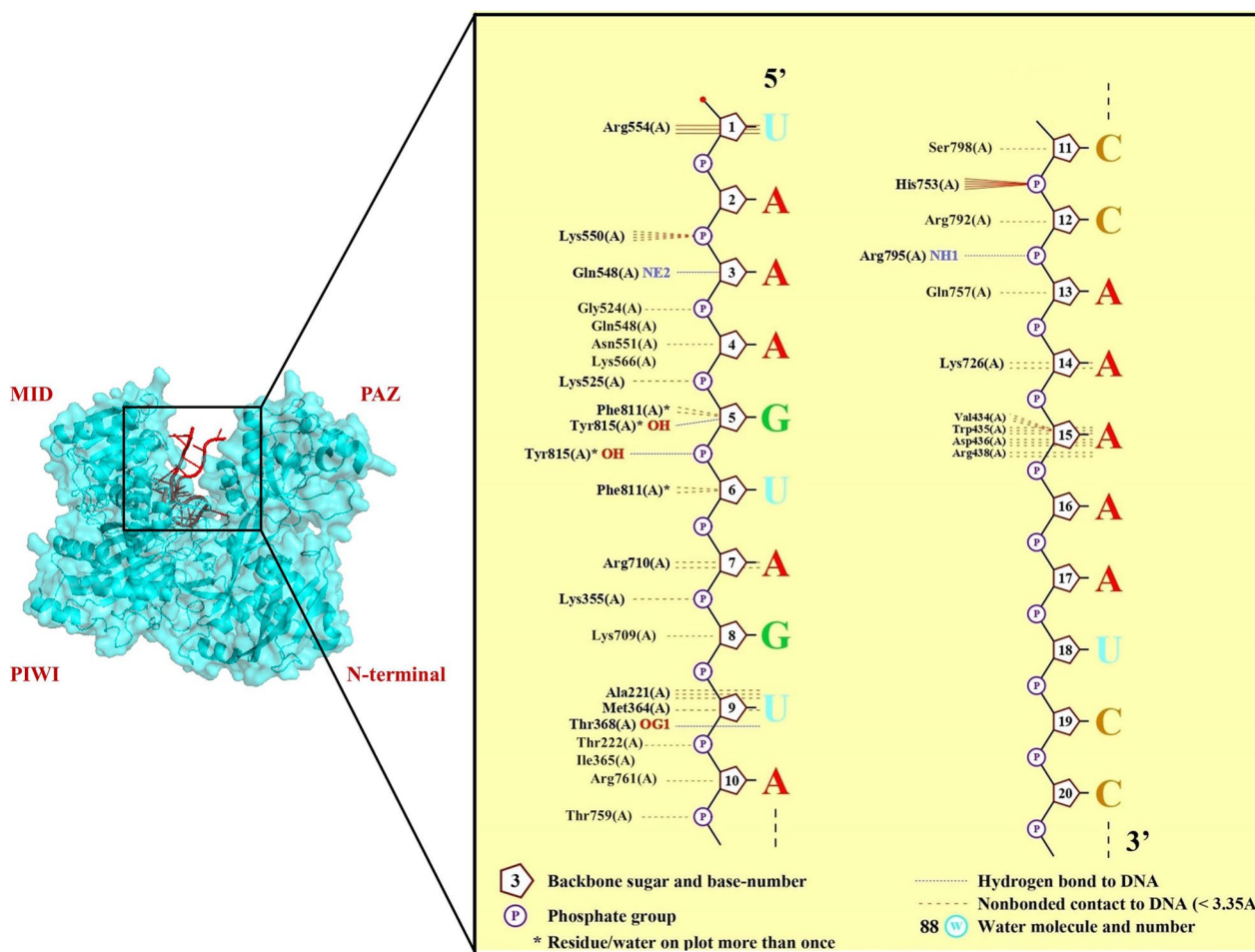
* Residues matched with previous studies; #Residues matched with control

cluster of novel mutations in the spike protein, on which the COVID-19 vaccines are based, indicating that the variant is partially resistant to pre-existing immunity [55].

Additionally, a recent study found that primary immunization with two doses of the ChAdOx1 nCoV-19 (AstraZeneca) or BNT162b2 (Pfizer-BioNTech) vaccine showed inadequate protection against the omicron variant [58]. Moreover, after either the ChAdOx1 nCoV-19 or BNT162b2 primary course, a booster dose of BNT162b2 or mRNA-1273 (Moderna) significantly improved protection, but this protection diminished with time [58]. It is also not negligible that the original strain of SARS-CoV-2 has an R0 of 2.5, while the delta variant (B.1.617.2) has an R0 of under 7 and now the omicron variant outcompetes this number with an R0 of 10 [55]. So, as a result of the strong transmission and immune evasion of the continuing Omicron variety, therapeutic research is critical to halt the spread of the fifth wave of the pandemic. Thus, siRNA, the next-generation therapy, must be effective in this scenario, which is why it is the focus of our research.

Here, in this study, first of all we have identified the possible siRNAs (21nt+2nt overhang region) with possible targets from the CDS of spike protein of omicron variant. This was done using the siDirect website, which conducts the work in three steps: highly functional siRNA selection, seed-dependent off-target effects minimization, and near-perfect matched genes deletion. The siDirect web server initially predicted 702 potential siRNA's with target from the CDS of spike protein. However, we sorted this 702 siRNAs to only 17 by selecting both combined U,R A method (Ui-Tei, Renold and Amargui-oui rules) and selecting those siRNA's whose seed-target duplex Tm is under 10 °C. Generally, siRNA's thermodynamic stability or seed-target duplex Tm under 21.5 °C is considered a benchmark as this minimizes the off-target effects [33]. This assured that our siRNAs are distinctive and have a low off-target binding rate.

The GC content of siRNA duplexes is one of the essential criteria for siRNA effectiveness, and GC content has an antagonistic connection with siRNA function [35]. When the content of GC is too high, the RISC complex-related helicase may take longer to unwind the



siRNA duplex. On the other hand, the smaller GC content may limit the efficacy of target mRNA identification and hybridization. It is, thus, recommended to pick siRNA sequences with low GC content (between 30 and 52%) [47]. In our study, we evaluated the siRNAs for GC content and eliminated the siRNAs whose GC content is fewer than 30%. This subsequently sorted our 17 siRNA’s to 5.

According to previous research, an RNA molecule should have the highest free energy of folding [59]. The formation of secondary structure in siRNA molecules owing to lower folding free energy may prevent target cleavage by RISC complex. As a result, it is critical to calculate the potential secondary structure and free energy of folding. SiRNA molecules with positive free energy of folding of the guide strands may have more access to the target and have a greater chance of interacting with it, resulting in successful gene silencing [60]. In our study, out of 5 siRNA’s two siRNA’s showed positive

free energy after analysis of secondary structure by RNA structure webserver. We then selected these two siRNAs (S2 & S10) for further analysis.

Because RNAi efficacy is highly connected with the binding energies of siRNAs to their respective target mRNAs, the free energy of binding with the target (i.e., computational RNA–RNA interaction) is another significant metric [61]. Lower binding energy suggests a stronger interaction, and hence an increased likelihood of inhibiting the target. After analysis, our final two siRNAs showed more negative free energy of binding (–30.2 for S2 and –28.4 for S10).

Additionally, the higher values of Tm (Cp) and Tm suggest that the siRNAs are more effective (Conc). The heat capacity plot indicates the Cp as a function of temperature, and when the Cp is a function of Tm, it is expressed as TmCp. Similar to a concentration plot, the mole fractions plotted as a function of temperature is represented by Tm (Conc). The concentration of the double-stranded

first of all we predicted the 3D structure of Ago2 protein with Robetta homology modeling web server. Afterward, we refined the modeled protein with Galaxy refine web-server. The resulting model had 99.062% residues laid in the highly preferred observation analyzed by Ramachandran plot analysis. Subsequently, we also modeled the 3D structure of our final two siRNAs with Mfold and RNA Composer webserver.

After modeling, we used the HDOCK website to dock our candidate siRNAs with human Ago2 protein. The docking complexes were downloaded from the web server and manually evaluated to determine the best-docked complex based on the docking score, the visual likeness of the complex to the 4Z4D structural composite, and the positioning of siRNA in the same binding pocket of 4Z4D. This analysis showed that model no 5 of S2 and model no 1 of S10 bind in the same pocket of the Ago2 protein and resembled the 4Z4D siRNA-Ago2 complex. Finally, we selected these modeled complexes for further RNA–protein interaction analysis.

The RNA–protein interaction of the S2, S10 and Ago2 complex showed that this siRNA binds in the same pocket of Ago2 spanning between the L1, PAZ, PIWI and Mid domain. However, both of these siRNA's strongly anchored in the PIWI domain of Ago2 wherever none of the siRNA's docked with the N-terminal site of Ago2. The docking score also revealed strong binding affinity with a score of -300.17 for S2 and -350.23 for the S10 molecule. However, it is clearly visible that S10 molecule outcompetes the S2 with greater binding affinity. The binding residues are also found better in the S10-Ago2 complex compared to the S2-Ago2 complex. We also analyzed the binding residues with a previous experimental analysis and found similarities for both of these complexes. For S2 molecule, interacting residues that found similarities with previous studies are ALA221, Thr222, ILE365, GLY524, ASN551, LYS709, ARG710, GLN757, THR759, ARG76, ARG792, SER798 [70–72]. And, for S10 molecule, residues that were found to be similar to previously reported residues are SER220, ALA221, THR222, ARG351, ILE365, LYS709, ARG710, GLN757, THR759, ARG761, TYR792, SER798, TYR804 [70–72]. However, both of these complex shares some common residues, e.g., ALA221, Thr222, ILE365, LYS709, ARG710, GLN757, THR759, ARG792, SER798. These residues are also found in previously reported experimental studies [70]. So, it can be stated that these residues are conserved for binding the siRNA's with human Ago2 protein.

From this structural perspective, though the S10-Ago2 complex showed better binding efficacy than S2-Ago2 complex, S2 molecules is 100% conserved against all 59 sub-lineages of the omicron variant. S10 molecule is not 100% conserved; however, S2 molecule could be

performed as 100% conserved against all sub-lineages of the omicron variant. A study revealed that small RNAs with an inaccurate match to native mRNA can also suppress translation [73]. In RISC-mediated RNA degradation, we know that a 21-base pair RNA duplex that matches perfectly an endogenous target mRNA selectively degrades the mRNA and reduces gene expression in mammalian tissue culture cells or viral cells. A study found that a mismatched RNA (up to 3–4 nt) directed to a particular spot in an endogenous gene's coding sequence can effectively inhibit gene expression by suppressing translation [73]. In our study, the target site of S10 molecule is 96.61% (out of 59, 57 is conserved), so we analyzed the mismatch pattern of the rest two sub-lineages with our target. This result revealed that these mismatches are due to one base alteration (Additional file 5: Fig. S4). So, according to the previous study, we can account for the fact that S10 molecules can affect the expression of spike protein of omicron variant through translational repression. This further revealed that S10 molecule can outperform S2 molecule in RNAi activity.

However, various obstacles, such as siRNA instability, limited cellular absorption, and the absence of a trustworthy delivery pathway, could pose difficulties for siRNAs' therapeutic potential for targeted gene silencing [74]. For effective gene therapy, a suitable promoter-controlled vector can help deliver therapeutic genes to the targeted cell [75]. Vector-based siRNA in plasmid form can also be used to target targeted genes within a given cell line to examine the potency of a newly created siRNA [76]. In our study, we have just identified the possible siRNA molecules for RNAi activity in the spike protein of the omicron variant. Further vector-based in-vitro research is needed to test our proposed two siRNAs. Various research groups have also proposed a similar RNAi treatment strategy for COVID 19 since the pandemic began [71, 72, 77]. However, no study is done yet on RNAi-mediated gene silencing against the omicron variant. Several pharmaceutical companies, including Siranomics, Vir Biotechnology, and OilX Pharmaceuticals, have discovered several SARS-CoV-2 RNAi targets and related siRNA agents. We hope our research will contribute to this landscape well [78]. Finally, the discovery of this siRNA therapeutic approach could be a potential alternative to traditional vaccine design in slowing down the COVID-19 pandemic.

Conclusion

RNAi treatment is a novel method for creating a variety of possible siRNA molecules for the post-transcriptional gene silencing of key genes in diverse biological organisms. The current study identified two single possible

siRNA molecule as an effective option for inhibiting the expression of spike protein in the omicron variant of Sars-cov-2 virus. We have specifically targeted the recent omicron offshoots which are now proliferating around the world, e.g., BA.4, BA.5, BQ.1, BQ1.1. and XBB. In the fight against viral infection, these two synthetic compounds might be exploited as innovative antiviral therapy, providing a foundation for academics and the pharmaceutical sector to create antiviral medicines at the genome level.

Abbreviations

| | |
|------------|--|
| SARS-CoV-2 | Severe acute respiratory syndrome coronavirus-2 |
| WHO | World Health Organization |
| ORF | Open reading frame |
| hACE2 | Human angiotensin converting enzyme 2 |
| RNAi | RNA interference |
| PTGS | Post-translational gene silencing |
| RISC | RNA-induced silencing complex |
| GISAID | Global Initiative for Sharing all Influenza Data |
| CDS | Coding sequence |
| Tm | Seed duplex's melting temperature |
| Cp | Ensemble heat capacity |
| UTR | 3' Untranslated region |
| GC | Guanine–cytosine content |
| RISC | RNA-induced silencing complex |
| VOC | Variant of concern |

Supplementary Information

The online version contains supplementary material available at <https://doi.org/10.1186/s43094-023-00510-3>.

Additional file 1. Table S1 GC content analysis of the predicted siRNA's.

Table S2: RNA Structure webserver prediction of free energy of folding of the predicted siRNA's.

Additional file 2. Figure S1: Heat capacity and concentration plot analysis of putative siRNA's. For siRNA molecule S2, the TmConc and TmCp is showed in (S2). For siRNA molecule S10, TmConc and TmCp is denoted in (S10).

Additional file 3. Figure S2: Conservancy analysis of the target RNA of the final two siRNA's against 59 omicron sub-lineages of SARS-CoV-2 virus. Target S2 molecule showed 100% conservancy whereas target of S10 molecule calculated 96% conservancy. Both of the consensus target sequences (denoted as black box) of siRNA molecule S2 and S10 is found 100% conserved in variant BA.4 and BA.5 (Shown in red box).

Additional file 4. Figure S3: Phylogenetic tree analysis of the 59 sub-lineages for spike protein of the SARS-CoV-2 omicron variant. A few number of lineages Showed significant divergence after tree analysis (bootstrap value>0.74).

Additional file 5. Figure S4: Mismatch pattern analysis of the target of siRNA S10 molecule with two sub-lineages (B.1.1.529 and BA.1). The mismatched is found for only one residual changes. Rest of the all sublineages found matched with the target.

Acknowledgments

No specific acknowledgment

Author contributions

RI contributed to conceptualization, project administration, supervision, writing. AS contributed to writing—review and editing. NF contributed to writing—review and editing. MRU contributed to writing—review and editing. MA contributed to writing—review and editing. MMRS contributed to

data curation, investigation, former analysis, writing. AS contributed to data curation, writing—review and editing, visualization. KJA contributed to review and editing.

Funding

No specific grant was received for this study.

Availability of data and materials

All data supporting the findings of this study are available within the article and its supplementary materials.

Declarations

Ethics approval and consent to participate

Ethical approval not required.

Consent for publication

All authors have consent to publish this manuscript.

Competing interests

The authors declare that they have no competing interests.

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Received: 30 January 2023 Accepted: 11 July 2023

Published online: 02 August 2023

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