

## Original Research Article

### **Computational Identification of YQG-like Small-Molecule Inhibitors Targeting TIM-3 for Cancer Immunotherapy**

#### **Abstract**

**Aims:** To identify potential small-molecule inhibitors of T-cell immunoglobulin and mucin domain-3 (TIM-3) through computational approaches and evaluate their interactions, stability, and structural dynamics.

**Study Design:** A computational drug discovery study utilizing virtual screening, molecular docking, and molecular dynamics simulations.

**Methodology:** A pharmacophore/similarity search was conducted using the PUBCHEM database, followed by molecular docking simulations to identify compounds with favorable binding properties to TIM-3. Three top-performing compounds (CID\_146311758-TIM3, CID\_164628526\_TIM-3, and CID\_146301996-TIM3) were analyzed further using molecular dynamics simulations to assess their binding stability, structural compactness, and hydrogen bond interactions. The study was conducted at the Department of Chemistry and Chemical Engineering, University of New Haven, USA, between January and December 2024.

**Results:** CID\_164628526\_TIM-3 displayed stable binding (-8.6 kcal/mol), minimal fluctuations, and a compact structure closely resembling the reference compound YQG (-8.3 kcal/mol). CID\_146311758-TIM3 (-8.6 kcal/mol) and CID\_146301996-TIM3 (-8.4 kcal/mol) showed higher flexibility and fluctuations. Hydrogen bond analysis indicated that CID\_164628526\_TIM-3 formed fewer bonds on average. These findings suggest that CID\_164628526\_TIM-3 is a promising candidate for further investigation.

**Conclusion:** This study highlights the potential of computational approaches to identify small-molecule inhibitors for TIM-3. CID\_164628526\_TIM-3 demonstrated superior binding energy (-8.6 kcal/mol) and stable interactions, closely resembling the reference compound YQG in structural integrity and hence providing a foundation for developing novel therapeutic agents targeting TIM-3 to enhance immune responses against tumors.

**Keywords:** Computer-aided drug design, TIM-3, Cancer immunotherapy, Molecular docking, Molecular dynamics simulation.

## 1.0 Introduction

Over the course of the past decade, there has been a remarkable transformation in the field of cancer treatment as immunotherapy has emerged as a groundbreaking approach (Baxevanis et al., 2009; Morrison et al., 2018). By harnessing the remarkable capabilities of the immune system to target and eliminate cancer cells, immunotherapy offers a highly promising pathway towards long-lasting and potentially curative responses across a wide range of cancer types (Abbott and Ustoyev, 2019). Central to the success of immunotherapy is the regulation of T-cell activation and effector functions, which rely on a delicate balance of co-stimulatory and co-inhibitory signals (Yu et al., 2019). Among these regulatory factors, immune checkpoints such as programmed cell death protein 1 (PD-1), cytotoxic T-lymphocyte-associated antigen 4 (CTLA-4), and T-cell immunoglobulin and mucin-domain containing-3 (TIM-3) play critical roles in preserving immune homeostasis and preventing autoimmunity (Abbott and Ustoyev, 2019; Yu et al., 2019).

TIM-3, a transmembrane protein expressed on select immune cells, has emerged as a pivotal immune checkpoint receptor in the realm of cancer immunology. Initially identified as a negative regulator of immune responses, particularly those of the Th1 type, TIM-3 has now been recognized for its involvement in tumor immunity and the establishment of immunosuppressive environments (O'Donnell et al., 2019). The presence of TIM-3 extends beyond a single immune cell type, with its expression observed on various key players such as CD4<sup>+</sup> and CD8<sup>+</sup> T-cells, regulatory T-cells (Tregs), natural killer (NK) cells, dendritic cells (DCs), and myeloid-derived suppressor cells (MDSCs) (Saleh and Elkord, 2020). Notably, within the tumor microenvironment, TIM-3 expression is heightened, suggesting its significant role in driving tumor-induced immune dysfunction. Activation of TIM-3 entails a complex series of interactions with its ligands, predominantly galectin-9 (Gal-9), phosphatidylserine (PtdSer), and

carcinoembryonic antigen-related cell adhesion molecule 1 (CEACAM1) (Zhang et al., 2016). This activation cascade leads to the recruitment of downstream signaling molecules, including Src homology 2 domain-containing protein tyrosine phosphatase-1 (SHP-1) and Bat3, culminating in the suppression of T-cell function. Evidence is mounting to suggest cross-talk between TIM-3 and other immune checkpoints, such as PD-1, CTLA-4, lymphocyte activation gene 3 (LAG-3), and T-cell immunoreceptor with Ig and ITIM domains (TIGIT) (Shi et al., 2021). These interactions underscore the complexity of immune checkpoint regulation and open up avenues for potential synergistic targeting approaches. The presence of elevated TIM-3 expression within tumor-infiltrating lymphocytes (TILs) has been correlated with unfavorable prognoses across multiple cancer types (Vlaming et al., 2022). Such high TIM-3 expression has been linked to more aggressive tumor characteristics, increased propensity for metastasis, and diminished overall survival rates, thereby highlighting its potential as a valuable prognostic biomarker (Qin et al., 2020). Additionally, TIM-3 expression has been put forward as a promising predictive biomarker for assessing immunotherapy response. Studies have demonstrated that tumors exhibiting heightened levels of TIM-3 expression tend to be less responsive to immune checkpoint blockade, suggesting the potential utility of TIM-3 in tailoring treatment strategies to individual patients (Das et al., 2017; Qin et al., 2020). Given the escalating importance of TIM-3 as a biomarker, ongoing efforts are dedicated to the development of robust and standardized techniques for detecting TIM-3.

The therapeutic blockade of TIM-3 utilizing antibodies has exhibited noteworthy achievements in the treatment of various human malignancies, and ongoing clinical trials further substantiate its potential (Acharya et al., 2020; Liikanen et al., 2022). It has been reported that the primary mode of action for anti-TIM-3 antibodies lies in their ability to impede the interaction between TIM-3 and PtdSer as well as CEACAM1 (Sebatos-Peyton et al., 2017). However, one limitation associated with antibodies is their relatively limited capacity to penetrate tumors effectively (Thurber et al., 2008). As a result, substantial efforts have been dedicated to the development of small molecule inhibitors targeting TIM-3 (Wu et al., 2023).

To date, only a single peptide and a solitary small molecule inhibitor have been reported as viable options in targeting TIM-3 (Wu et al., 2023). The small molecule inhibitor known as CA-327 demonstrates the unique capability of targeting both TIM-3 and PD-L1, thereby exhibiting a

certain degree of anti-tumor efficacy (Wu et al., 2023). Nonetheless, the structural details of CA-327 have not been publicly disclosed, and a comprehensive understanding of the precise mechanisms through which it exerts its anti-tumor activity remains incomplete (Testa et al., 2020). Consequently, the exploration and identification of novel small molecule inhibitors that specifically target TIM-3 assume utmost significance in the realm of cancer treatment.

The quest for novel small molecule inhibitors represents a critical step forward in combating cancer by effectively targeting TIM-3. By expanding our arsenal of therapeutic options, we can potentially overcome the limitations associated with antibody-based treatments and advance towards more tailored and efficacious approaches for TIM-3 blockade (Wu et al., 2023).

In this study, the aim of our research is to investigate the therapeutic potential of targeting TIM-3 in cancer treatment. By employing computational chemistry and drug discovery approaches, our goal is to identify novel small molecule inhibitors that can effectively bind to TIM-3, modulate its activity, and disrupt the immunosuppressive tumor microenvironment. As a starting point for our virtual screening campaign, we performed a pharmacophore/similarity search utilizing the PUBCHEM database to identify compounds similar to YQG, a co-crystal inhibitor of TIM-3. Next, to prioritize and select compounds with favorable binding energies and key molecular interactions, we conduct molecular docking simulations to explore the binding interactions between the identified compounds and the active site of TIM-3. Finally, molecular dynamics simulations was performed to investigate the conformational dynamics, stability, and intermolecular interactions of the selected compounds with TIM-3 over an extended period of time.

## **2.0 Materials and Methods**

### **2.1 Pharmacophore/Similarity Search**

For the purpose of identifying compounds that bear resemblance to YQG, which happens to be a co-crystal inhibitor of TIM-3 (Rietz et al., 2021), we carefully conducted a pharmacophore/similarity search employing the vast resources of the PUBCHEM database (Kim et al., 2021). This particular approach entailed scrutinizing an extensive collection of compounds in order to uncover those possessing comparable chemical attributes and structural traits akin to

YQG. From this comprehensive search, we curated a subset of 143 compounds, which were subsequently earmarked for subsequent in-depth analysis and investigation.

## **2.2 Structure-based Virtual screening**

We used the cutting-edge AutodockVina program, which is known for its ability in this domain, to carry out the careful docking simulation (Trott and Olson, 2010). The protein structure of TIM-3, which served as the main component of our study, was sourced from Protein Data Bank (PDB) under the unique identification 7m41 (Rietz et al., 2021). By carefully identifying and defining the protein's active site, we focused our attention on a set of residues, including Asp53, Val54, Trp57, Ser59, Arg60, Tyr61, Trp62, Phe67, and Arg68, which together form TIM-3's main pocket (Rietz et al., 2021). Careful attention was paid to defining the size of the grid box in order to precisely map the spatial properties of the protein active site. This grid box's dimensions were specified with exactitude to be 60 units along the x-axis, 50 units along the y-axis, and 40 units along the z-axis. This grid box's center was carefully determined to be at (-19.079, -37.352, 5.951), which pinpointed the focus of our docking exploration. In order to achieve the accurate modeling of the intermolecular interactions within this simulated environment, a spacing of 0.375 angstroms was purposefully chosen. Our work moved on to the important phase of docking simulations once the active region of TIM-3 was clearly identified. The findings of the prior pharmacophore search yielded 143 compounds, which were carefully chosen and then subjected to this comprehensive computational analysis. By carefully examining the interactions and probable binding mechanisms between these substances and the targeted active site of TIM-3, we were able to get crucial knowledge about prospective inhibitory candidates deserving of further study. Finally, Biovia discovery studio software (<https://www.3ds.com/products/biovia/discovery-studio>) were used to visualize and present the docking interactions in 2D and 3D formats.

## **2.3 Molecular Dynamics Simulation**

The comprehensive exploration of our research involved not only the pharmacophore/similarity search and docking simulations but also encompassed the pivotal step of conducting molecular dynamics (MD) simulations utilizing the cutting-edge GROMACS 2022.3 software (Van Der Spoel et al., 2005). Our focus was specifically directed towards four complexes of immense interest: CID\_146311758-TIM3, CID\_164628526\_TIM-3, CID\_146301996-TIM3, and YQG-

TIM3. To ensure the accurate representation of the intricate interplay between ligands and the protein, we employed the CHARMM36 force field to effectively parameterize both components within the system (Vanommeslaeghe et al., 2010). By carefully accounting for the diverse molecular properties, we established a foundation for a rigorous investigation into the conformational dynamics and stability of the ligand-protein interactions.

The solvation step involved surrounding the complexes under investigation, namely CID\_146311758-TIM3, CID\_164628526\_TIM3, CID\_146301996-TIM3, and YQG-TIM3, with an appropriate solvent environment. This solvent, typically water, was carefully chosen to mimic the physiological conditions relevant to the study. By solvating the system, we created a realistic environment that allowed for the accurate representation of the interactions between the compounds and the surrounding solvent molecules. Following solvation, the system underwent an ionization step. Here, appropriate ions (2 sodium ions) were added to maintain the overall system neutrality and to simulate the desired ionic strength. This step ensured the preservation of electrostatic balance within the system, crucial for simulating realistic physiological conditions. Next, the system underwent a minimization process of 5000 nsteps, where steepest descent energy optimization algorithms were employed to relax the initial structure and eliminate any steric clashes or unfavorable interactions. This step involved iteratively adjusting the positions of atoms to reach a more stable energy state, ensuring a starting point for the subsequent MD simulations that was free from significant structural distortions or artifacts. After minimization, the equilibration step was performed to gradually adjust the system from the minimized state to the desired simulation conditions. This involved allowing the system to reach equilibrium by controlling parameters such as temperature and pressure. The temperature coupling (tcoupl) and pressure coupling (pcoupl) were set as V-rescale and Berendsen respectively. During this process, the system was subjected to a series of restrained or constrained dynamics, gradually releasing these restraints to allow for more freedom of motion and achieving a stable equilibrium state. These steps—solvation, ionization, minimization, and equilibration—were integral in setting up the system for the subsequent MD simulations. By meticulously following these protocol steps, we ensured that the system was well-prepared, stable, and representative of the physiological conditions of interest. The MD simulations were then conducted over a production run spanning a duration of 100 nanoseconds (ns) for each of the aforementioned complexes. This extended timescale allowed us to explore and elucidate the subtle nuances of the ligand-protein

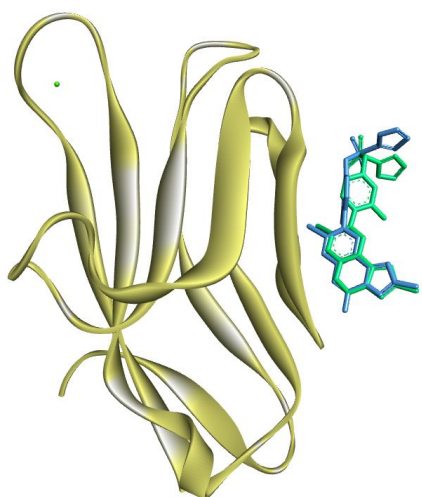
interactions, shedding light on their binding modes, calculating binding energies, and carefully monitoring the structural stability and dynamics of these complexes throughout the simulation trajectories.

### **3.0 Results and Discussion**

#### **3.1 Molecular docking-based virtual screening**

In our docking-based virtual screening method, we initiated the process by conducting a redocking experiment. The goal was to assess the accuracy and reliability of our docking protocol by comparing the binding mode of the redocked YQG compound with the experimentally determined co-crystal pose of YQG (Rietz et al., 2021). Encouragingly, our redocking results indicated a close resemblance between the binding modes of the redocked YQG compound and the reference co-crystal structure, thereby validating the efficacy of our docking protocol (Figure 1) (Adelusi et al., 2021). The calculated RMSD between the experimental and redocked poses was estimated to be approximately 1.3 Å, indicating a high degree of alignment. This RMSD value falls well below the widely accepted threshold of 2.0 Å, confirming the reliability of the docking procedure in reproducing the experimental binding pose. Building upon this validation, we proceeded with docking simulations involving 143 pharmacophores derived from the YQG compound. Our aim was to identify potential compounds that exhibit superior binding energy in comparison to the reference molecule, YQG, which possessed a binding energy of -8.3 kcal/mol (Figure 2). The docking results unveiled three noteworthy compounds, namely CID\_146311758-TIM3, CID\_164628526\_TIM-3, and CID\_146301996-TIM3, all of which demonstrated more favorable binding energies (-8.6 kcal/mol, -8.6 kcal/mol, and -8.4 kcal/mol, respectively) when compared to YQG (Figure 2, Supplementary Table 1). Further analysis of the binding interactions shed light on the intricate molecular associations between these compounds and the active pocket of TIM3 (Rietz et al., 2021). CID\_146311758-TIM3 was found to engage in four hydrogen bonds with key residues, namely Asp53, Arg60, and Arg68 (Figure 3). Additionally, pi-pi stacking interactions were observed with Trp57 and Trp62, along with pi-sigma bonding with Val54, pi-sulfur interaction with Phe67, and a carbon-hydrogen bond involving Ser59. As for CID\_164628526\_TIM-3, its binding interaction profile included pi-pi stacking with Trp57 and Trp62, pi-alkyl bonding with

Val54 and Phe67, and a pi-pi T-shaped interaction with Phe67 (Figure 3). Lastly, CID\_146301996-TIM3 demonstrated four hydrogen bond interactions with Asp53, Arg60, and Arg68, although an unfavorable hydrogen bond was observed with Tyr61 (Figure 3). Additionally, pi-sulfur bonding with Phe67, pi-sigma interaction with Val54, pi-pi stacking interactions with Trp57 and Trp62, and a carbon-hydrogen bond with Ser59 contributed to its binding pattern. Overall, these interactions firmly establish the binding of the four identified compounds to the active pocket of TIM3 protein. Consequently, they hold considerable promise as potential drug candidates for the treatment of TIM3-associated cancer patients. The comprehensive molecular docking analysis, encompassing the validation of our docking protocol, the identification of superior binding energies, and the detailed characterization of key binding interactions, contributes to our understanding of the molecular mechanisms underlying the interactions between these compounds and TIM3.



*Figure 1: Superimposition of the redocked X-ray structure of YQG (green) with the experimental pose (blue).*



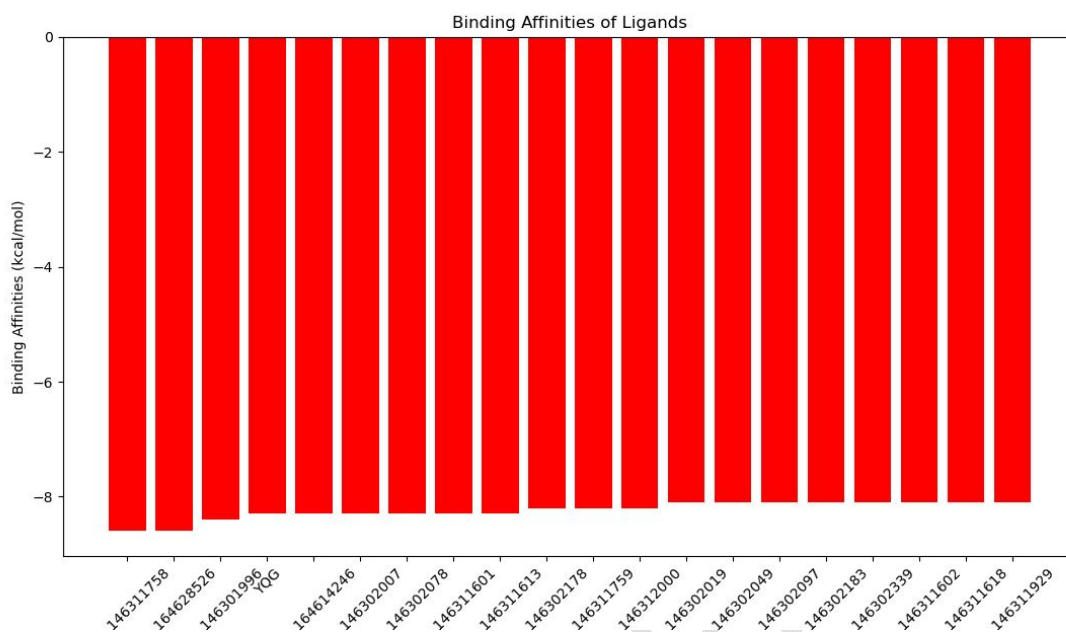
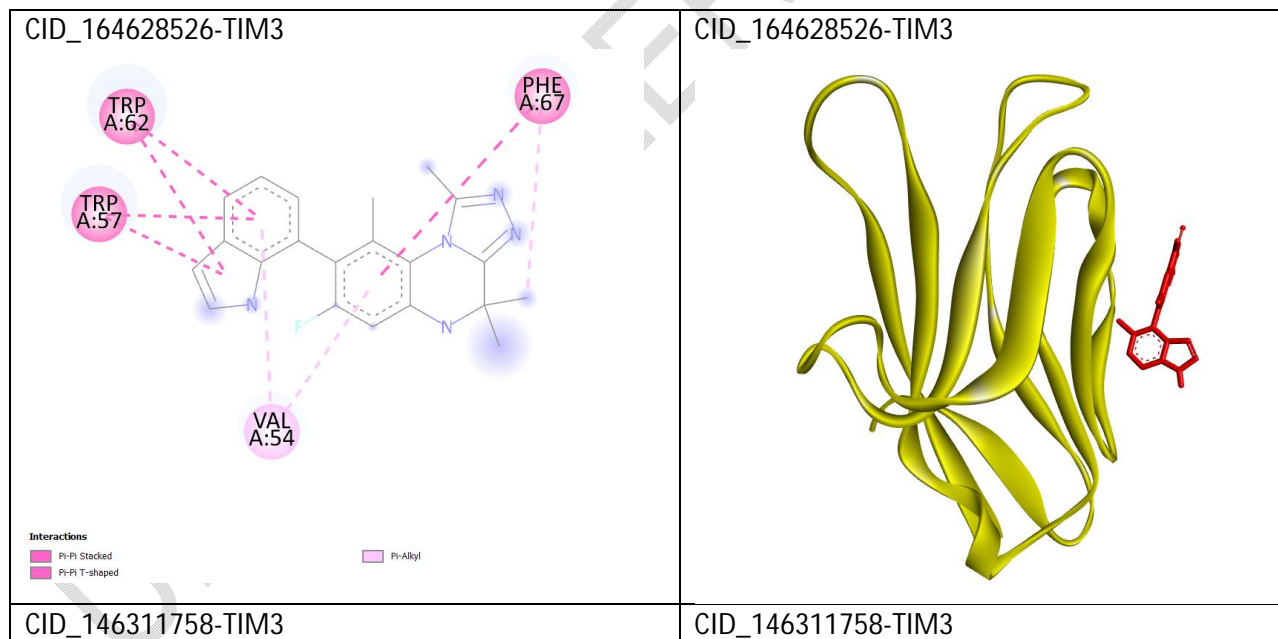


Figure 2: Top 20 scoring compounds from the Docking-based virtual screening.



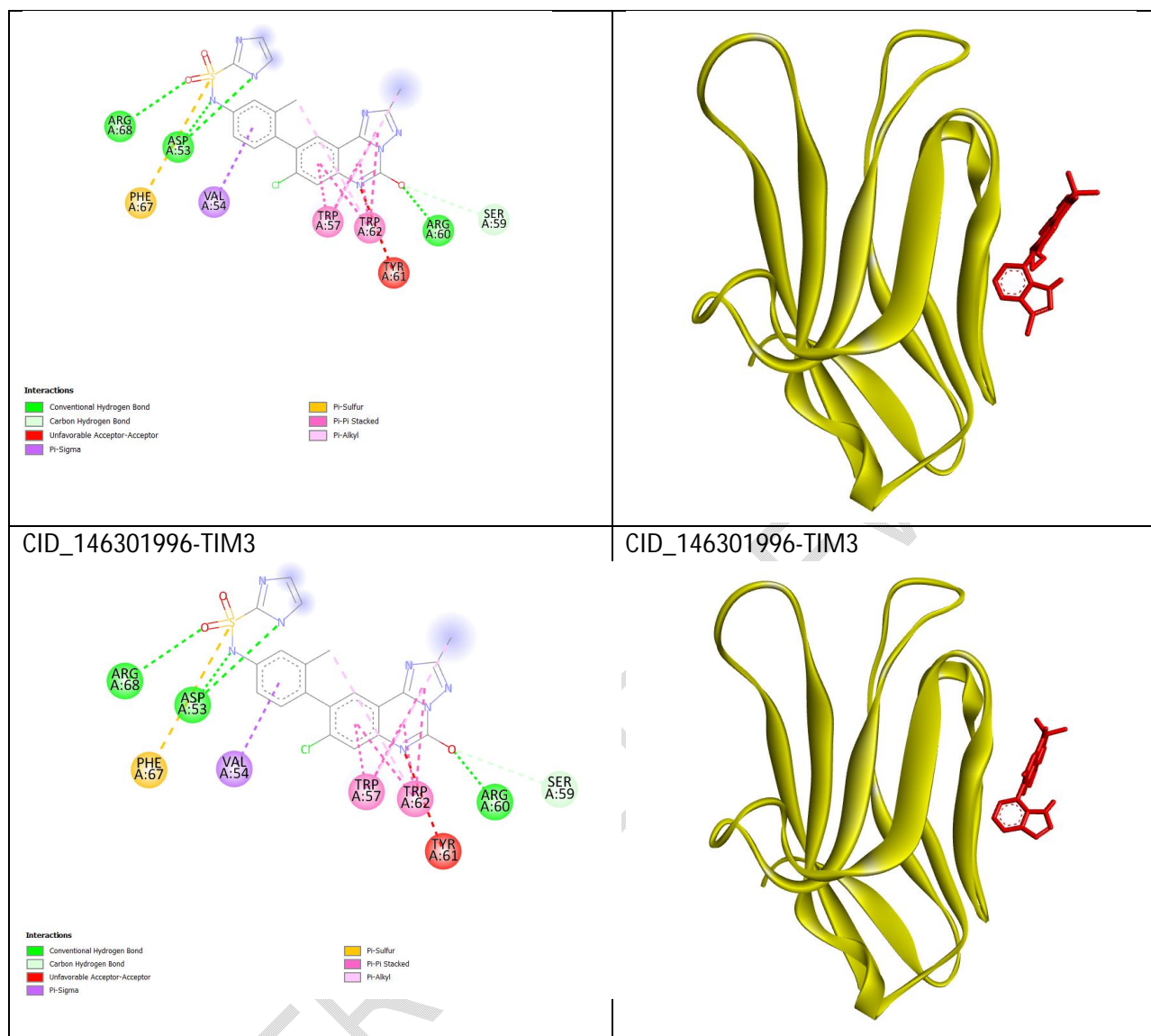


Figure 3: The 2D (Left) and 3D (Right) structures of the Top3 hit candidates in complex with the TIM3 receptor.

## 3.2 Molecular dynamics simulation

### 3.2.1 RMSD analysis

During the molecular dynamics simulation, the Root Mean Square Deviation (RMSD) was calculated to assess the stability and structural changes of the complexes (Ogunlana et al., 2022). The average RMSD values for the different complexes are as follows: CID\_146311758-TIM3 averaged 0.55 nm, CID\_164628526\_TIM-3 averaged 0.30 nm, CID\_146301996-TIM3 averaged 0.46 nm, and YQG-TIM3 averaged 0.23 nm (Table 1). The RMSD values provide insights into

the deviation of each complex's structure from the starting conformation (Adelusi et al., 2022). By analyzing the spectrum of each complex, it is evident that CID\_164628526\_TIM-3 exhibits minimal fluctuation throughout the entire simulation period and closely resembles the standard (YQG) in terms of the average RMSD (Figure 4). This indicates that CID\_164628526\_TIM-3 maintains a stable structure with relatively minor changes over time. In contrast, CID\_146311758-TIM3 and CID\_146301996-TIM3 do not maintain a consistently stable RMSD spectrum throughout the entire simulation (Figure 4). CID\_146311758-TIM3 shows maximal or abrupt fluctuation between 38 ns and 95 ns before the spectrum sharply decreases around 97 ns. Subsequently, it maintains minimal fluctuation until the end of the simulation. Similarly, CID\_146301996-TIM3 exhibits abrupt fluctuation starting around 50 ns, which continues throughout the entire 100 ns simulation period. This suggests that CID\_146301996-TIM3 undergoes more substantial structural changes and displays greater flexibility compared to the other complexes. The observed fluctuations in RMSD indicate variations in the conformations and structural integrity of the complexes over time (Oyedele et al., 2022). These findings highlight the importance of considering the stability and structural dynamics of the compounds when assessing their potential as drug candidates.

*Table 1: The statistical average mean values of the RMSD, RMSF, ROG and H-bond of the compounds (including the Standard) in complex with TIM3.*

Complexes	Average RMSD (nm)	Average RMSF (nm)	Average ROG (nm)	Average H-bond (nm)
CID_164628526_TIM-3	0.30	0.14	1.39	0.58
CID_146311758-TIM3	0.55	0.21	1.46	0.77
CID_146301996-TIM3	0.46	0.18	1.42	0.20
YQG-TIM3	0.23	0.14	1.38	2.28

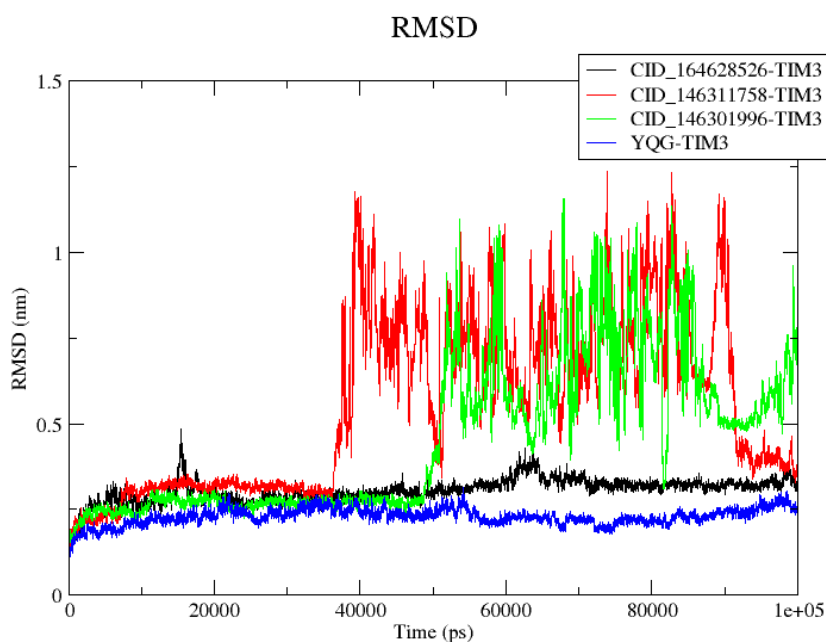


Figure 4: The RMSD spectrum of the top scoring compounds and YQG in complex with TIM3.

### 3.2.2 RMSF analysis

The Root Mean Square Fluctuation (RMSF) analysis provides insights into the flexibility and dynamic behavior of the complexes during the molecular dynamics simulation. By calculating the average RMSF values for each complex, we can assess the extent of atomic fluctuations within the protein structure (Ogunlana et al., 2023; Egbemhenghe et al., 2024). The average RMSF values for CID\_164628526\_TIM3, CID\_146311758-TIM3, CID\_146301996-TIM3, and YQG-TIM3 are 0.14 nm, 0.21 nm, 0.18 nm, and 0.14 nm, respectively (Table 1). Interestingly, both CID\_164628526\_TIM3 and YQG-TIM3 display the same average RMSF value of 0.14 nm. This suggests that these two complexes exhibit similar levels of atomic fluctuations within their respective protein structures (Figure 5). The consistency between the RMSF and RMSD results further supports the notion that CID\_164628526\_TIM3 is a promising hit candidate. The lower average RMSF value indicates that this compound experiences relatively less fluctuation and maintains a more stable protein structure during the simulation (Adelusi et al., 2022). This finding aligns with the observation from the RMSD analysis, where CID\_164628526\_TIM3

demonstrated minimal fluctuation and closely resembled the standard YQG-TIM3. Comparatively, CID\_146311758-TIM3 and CID\_146301996-TIM3 display higher average RMSF values of 0.21 nm and 0.18 nm, respectively (Table 1). These complexes exhibit relatively more atomic fluctuations, suggesting increased flexibility within their protein structures. The consistency between the RMSF and RMSD results reinforces the potential of CID\_164628526\_TIM-3 as a promising hit candidate. Its stable structure, combined with minimal fluctuations and low RMSD values, indicates a higher degree of conformational stability and structural integrity (Oyedele et al., 2023; Manica et al., 2025). By considering both the RMSF and RMSD analyses, we can confidently conclude that CID\_164628526\_TIM-3 stands out as the most promising hit candidate. Its stable structure, minimal fluctuations from RMSF and RMSD values make it a strong contender for further investigation and potential therapeutic applications.

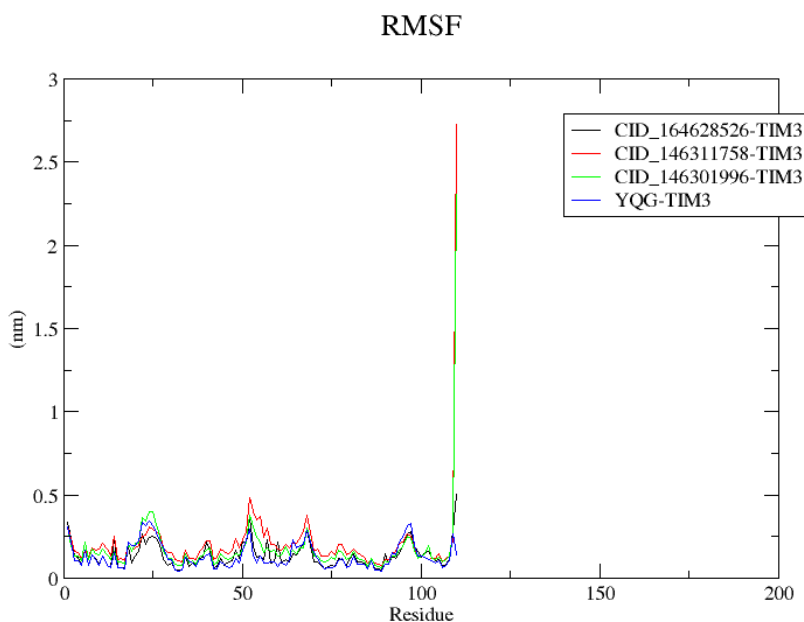


Figure 5: The RMSF spectrum of the top scoring compounds and YQG in complex with TIM3.

### 3.2.3 ROG

The Radius of Gyration (ROG) analysis provides valuable insights into the compactness and overall size of the protein structures within the complexes (Adelusi et al., 2022). By examining the average ROG values for each complex, we can assess their relative compactness and compare them to the reference YQG-TIM3. The average ROG values for CID\_164628526\_TIM-3, CID\_146311758-TIM3, and CID\_146301996-TIM3 are 1.39 nm, 1.46 nm, and 1.42 nm, respectively (Table 1). Comparing these values to the reference YQG-TIM3, which has an average ROG value of 1.38 nm, we can observe that all the complexes exhibit similar levels of compactness, as their ROG values are in close proximity. Again, the ROG results align with the findings from the RMSD and RMSF analyses, further reinforcing the reliability and consistency of our observations. The lower ROG value for CID\_164628526\_TIM3, the most promising hit candidate, indicates a slightly more compact structure compared to the other complexes. This suggests that CID\_164628526\_TIM3 may have a more stable and tightly folded protein conformation, potentially enhancing its binding affinity and stability within the active site of TIM-3 (Oyedele et al., 2022).

In addition to the average ROG values, it is important to consider the ROG spectrum and fluctuations exhibited by each complex during the simulation. By examining the ROG spectra, we can gain insights into the dynamic behavior and changes in the compactness of the protein structures over time (Adelusi et al., 2022). Upon analyzing the ROG spectra, it becomes evident that CID\_164628526\_TIM-3 and YQG-TIM3 exhibit remarkably similar patterns throughout the entire simulation period (Figure 6). This similarity suggests that these two complexes maintain consistent levels of compactness, indicating a relatively stable conformation and limited structural fluctuations. In contrast, CID\_146311758-TIM3 and CID\_146301996-TIM3 display more pronounced ROG fluctuations (Figure 6). After approximately 39 ns, CID\_146311758-TIM3 shows a notable increase in ROG values, indicating a temporary expansion or alteration in its protein structure. Similarly, CID\_146301996-TIM3 exhibits a significant increase in ROG values starting around 50 ns, suggesting a distinct change in its compactness. However, both complexes experience a subsequent decrease in ROG values around 90 ns, indicating a return to a more compact conformation. These observations parallel the findings from the RMSD analysis, where CID\_146311758-TIM3 and CID\_146301996-TIM3 demonstrated maximal fluctuations within their structures. The agreement between the ROG and RMSD results reinforces the consistency and reliability of our findings. Furthermore, the comparison of ROG values among

the complexes highlights the superior performance of CID\_164628526\_TIM-3. Its lower average ROG value, coupled with minimal fluctuations and closer resemblance to the reference YQG-TIM3, suggests a more stable and compact protein structure throughout the simulation (Boyenle et al., 2022). This stability may contribute to the potential of CID\_164628526\_TIM-3 as a promising hit candidate for further investigation and potential therapeutic development. By considering both the average ROG values and the dynamic ROG spectra, we gain a comprehensive understanding of the structural changes and compactness variations exhibited by the complexes. Our findings suggest that CID\_164628526\_TIM-3 demonstrates stable binding interactions and minimal conformational fluctuations, making it a promising therapeutic candidate for TIM-3 inhibition. Future in vitro studies will be critical to validate its effectiveness in reversing tumor immunosuppression, paving the way for preclinical development.

### Radius of gyration (total and around axes)

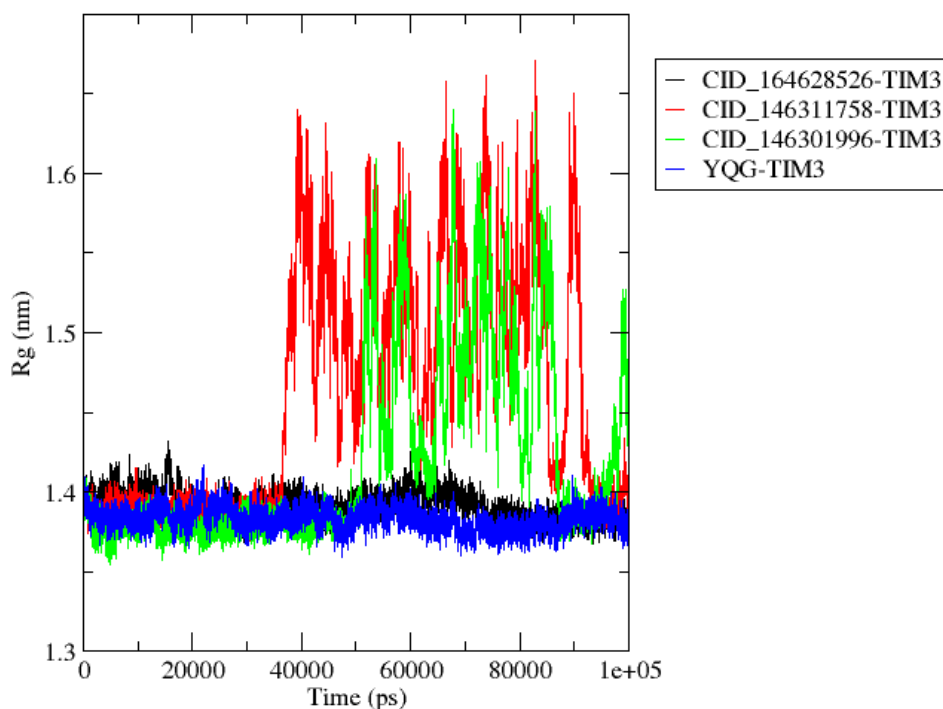


Figure 6: The ROG spectrum of the top scoring compounds and YQG in complex with TIM3.

### 3.2.4 Intermolecular H-bond

The analysis of intermolecular hydrogen bonds could provide valuable insights into the strength and frequency of these interactions between the ligands and the TIM-3 protein. By examining the average number of H-bonds formed by each complex, we can assess their potential for stable and specific binding interactions (Ayoub et al., 2014). The average number of H-bonds for CID\_164628526\_TIM-3, CID\_146311758-TIM3, and CID\_146301996-TIM3 were found to be 0.58, 0.77, and 0.20, respectively (Table 1 and Figure 7). In comparison, YQG-TIM3 exhibited an average of 2.28 H-bonds. The results indicate varying degrees of H-bond formation among the complexes. CID\_164628526\_TIM-3 displayed the lowest average number of H-bonds, indicating a relatively weaker tendency for H-bond formation with the TIM-3 protein. On the other hand, CID\_146311758-TIM3 had a slightly higher average number of H-bonds, suggesting a stronger potential for H-bond interactions (Figure 7). Interestingly, CID\_146301996-TIM3 showed the lowest average number of H-bonds among the three compounds, indicating a relatively weaker propensity for H-bond formation with TIM-3 (Adelusi et al., 2022). This observation suggests that CID\_146301996-TIM3 may rely on other types of interactions, such as hydrophobic interactions or electrostatic forces, for binding to the target protein. In contrast, YQG-TIM3 exhibited the highest average number of H-bonds, indicating a stronger capability to form these specific interactions with the TIM-3 protein. This finding is consistent with YQG being a known co-crystal inhibitor of TIM-3 and supports its role as a reference compound in our study (Rietz et al., 2021). The variation in H-bonding observed among the compounds underscores the importance of analyzing different molecular interactions in drug discovery. Although CID\_164628526\_TIM-3 forms fewer hydrogen bonds (0.58 on average) compared to YQG, its stable binding energy suggests a reliance on other interactions such as  $\pi$ - $\pi$  stacking and hydrophobic contacts, which enhance stability and specificity within the TIM-3 active site. Overall, the intermolecular H-bond analysis provides valuable information about the strength and nature of the interactions between the compounds and the TIM-3 protein (Adelusi et al., 2022).



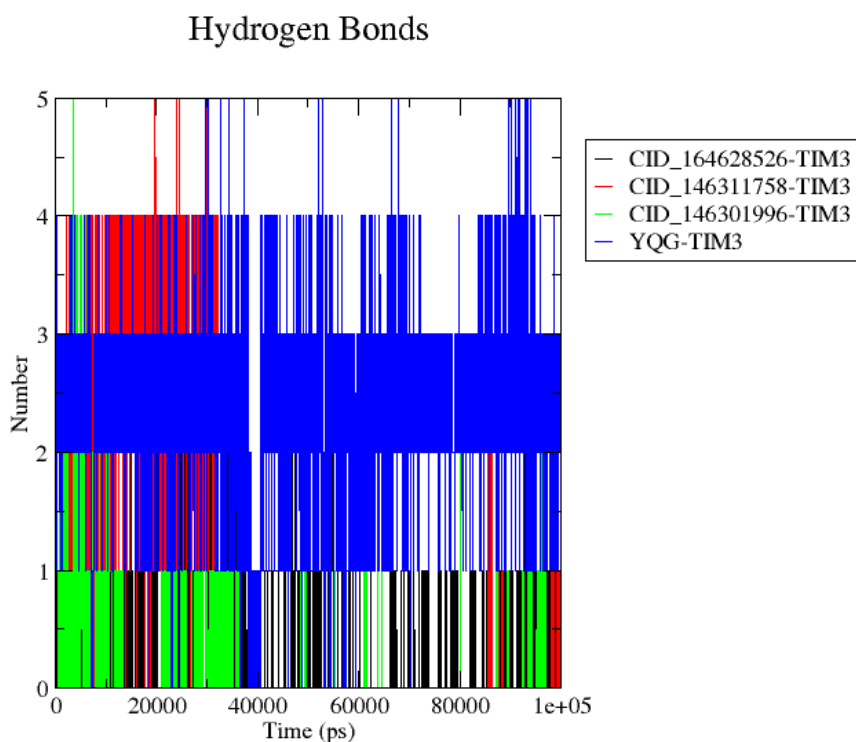


Figure 7: The Intermolecular H-bond spectrum of the top scoring compounds and YQG in complex with TIM3.

## Conclusion

In recent years, TIM-3 has emerged as a promising target for cancer immunotherapy due to its involvement in tumor immune evasion and suppression. Dysregulated TIM-3 signaling has been associated with immune dysfunction and tumor progression in multiple cancer types, making it an attractive target for therapeutic intervention. However, despite the potential of TIM-3 as a therapeutic target, the development of effective inhibitors remains a challenge. Traditional drug discovery approaches are often time-consuming and costly, requiring extensive screening of large compound libraries. In this context, computational methods such as molecular docking-based virtual screening and molecular dynamics simulations offer valuable tools for accelerating the identification and optimization of potential drug candidates.

In this study, our molecular docking-based virtual screening and molecular dynamics simulation analyses have provided valuable insights into the potential of compounds for targeting TIM-3 in cancer treatment. The docking results identified three compounds, namely CID\_146311758-TIM3, CID\_164628526\_TIM-3, and CID\_146301996-TIM3, which exhibited superior binding energies compared to the reference molecule, YQG. Detailed analysis of the binding interactions revealed the intricate molecular associations between these compounds and the active pocket of TIM-3. The molecular dynamics simulation further assessed the stability and dynamic behavior of the complexes through RMSD, RMSF, and ROG analyses. Among the compounds, CID\_164628526\_TIM-3 demonstrated minimal fluctuations, low RMSD and RMSF values, and a slightly more compact structure, closely resembling the reference YQG-TIM3. These findings suggest a higher degree of conformational stability and structural integrity for CID\_164628526\_TIM-3. Additionally, the intermolecular H-bond analysis highlighted varying degrees of H-bond formation among the compounds, with CID\_146311758-TIM3 exhibiting the strongest potential for these interactions. Collectively, these results suggest that CID\_164628526\_TIM-3 is a promising hit candidate for further investigation and potential therapeutic applications. As this work is purely *in silico research*, further optimization and design efforts can leverage the insights gained from this study to develop small molecule inhibitors with enhanced binding affinity and specificity for TIM-3, ultimately benefiting patients with TIM-3-associated cancers.

#### **Clinical trial number**

Not applicable.

Disclaimer (Artificial intelligence)

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- 2.
- 3.

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