Computational Insights into Breast Cancer Therapeutics: Auto-Docking Studies of

Symmetrical Azines with Breast Cancer Protein and Toxicity Predictions

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Abstract

Breast cancer is a prevalent malignancy affecting both women and men globally, arising from abnormal cell growth in breast tissues. Various risk factors, including age, genetics, hormonal influences, and lifestyle choices, contribute to its development. Early detection through screening methods like mammograms is pivotal for successful treatment. Treatment modalities encompass surgery, radiation, chemotherapy, hormone therapy, and targeted therapies, with outcomes improving due to heightened awareness and medical advancements. Auto-docking studies, employing computational simulations, are instrumental in drug discovery by predicting ligand-protein interactions. Breast Cancer Protein 2 (BCP2), represented by PDB code 2IOK, is pivotal in breast cancer pathogenesis, offering insights for targeted therapies. Here, we conducted auto-docking studies with four symmetrical azines on 2IOK, revealing azine one's superior binding affinity and toxicity predictions through ProTox-II software, highlighting potential therapeutic avenues in breast cancer research.

Key Words: Azine, Molecular Docking, Breast Cancer, Toxicity.

Introduction

Breast cancer is a type of cancer that forms in the cells of the breast. It is one of the most common cancers affecting women worldwide, but it can also occur in men, although much less frequently. Breast cancer can develop in different parts of the breast, including the milk ducts, the lobules that produce milk, or in the fatty tissue [1-3].

The exact cause of breast cancer is often unclear, but several factors can increase the risk, including age, family history, genetic mutations, hormonal factors, lifestyle choices, and environmental factors [4, 5].

Symptoms of breast cancer may include a lump or thickening in the breast or underarm area, changes in breast size or shape, skin changes on the breast, such as dimpling or puckering, nipple discharge other than breast milk, and nipple inversion [6-10].

Early detection through screening methods such as mammograms can greatly improve the chances of successful treatment. Treatment options for breast cancer may include surgery, radiation therapy, chemotherapy, hormone therapy, targeted therapy, or a combination of these approaches, depending on the type and stage of the cancer [11-15].

Awareness, early detection, and advancements in treatment have significantly improved outcomes for individuals diagnosed with breast cancer, but ongoing research and education efforts remain crucial in the fight against this disease.

Auto-docking studies involve computational simulations aimed at predicting the binding modes and affinities of small molecules or ligands with target proteins. This methodology is widely employed in drug discovery and design, as it helps in understanding the molecular interactions between ligands and their target binding sites. Auto-docking studies utilize algorithms and molecular modeling techniques to explore the conformational

space of ligands and predict their most favorable binding poses within the protein's binding site. These studies play a crucial role in virtual screening campaigns and lead optimization efforts, contributing to the development of novel therapeutics and bioactive compounds [16-19].

Breast cancer protein 2 (BCP2), also known as 2IOK, is a significant protein associated with breast cancer. The structure of this protein, represented by its PDB code 2IOK, provides crucial insights into its function and potential therapeutic targets. Understanding the structure and function of BCP2 is vital in unraveling the mechanisms underlying breast cancer development and progression. Researchers analyze 2IOK and related proteins to elucidate their roles in cell signaling, proliferation, and metastasis, aiming to develop targeted therapies for breast cancer patients [20, 21].

In this study, we selected four symmetrical azines that we had previously reported, and all four were docked with the breast cancer protein 2IOK. Among these azines, azine one exhibited superior binding affinity compared to the others. Additionally, the toxicity of all the azines was predicted using ProTox-II software.

Materials and Methods

Molecular docking

The protein data bank (PDB: 2IOK) contains Compound 1D and the human estrogen receptor alpha ligand-binding domain. Researchers investigated molecular docking using AutoDock 4.0 and an empirical grading scheme based on binding free energy. The AutoDock provides a wide range of stochastic search techniques. Our initial choice was to use the Lamarckian Genetic Algorithm (LGA). It utilises just the Genetic Algorithm to integrate both

local and global search (Solis and Wets algorithm). The 3D picture acquired from Discovery studio and the 2D image produced using the LigPlot.

Toxicity Prediction

We can cut down on preclinical drug development expenses, time, and animal testing by utilising in silico prediction techniques. ProTox-II includes machine-learning models covering hepatotoxicity, cytotoxicity, mutagenicity, immunotoxicity, adverse outcomes pathways (Tox21), and toxicity targets in addition to molecular similarity, pharmacophores, fragment probabilities, and other topics [22-24].

Results and Discussions

New compounds that could be used as inhibitors for a range of diseases are continuously being developed and synthesised by our research team [25-27]. As part of our ongoing study, we are currently utilising clinical techniques against humans in an effort to uncover a unique, potent chemical that has a favourable effect on multiple enzymes. The four chemicals were chosen, and similar compounds were even created in a wet lab and previously documented by our team of researchers. The structure of the four azine compounds is shown in Figure 1.

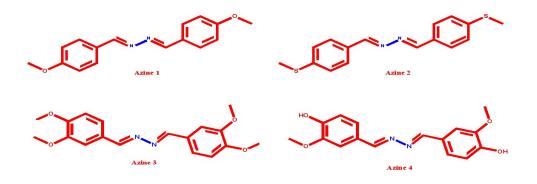


Figure 1. Structure of the azine 1-4

Docking Studies

Docking tests were performed to examine the active sites of the medicinal drugs against the breast cancer protein (2IOK) (Table-1).

		2IOK		
	Azine 1	Azine 2	Azine 3	Azine 4
Binding energy	-6.27	-6.06	-5.56	-5.58
Ligand efficiency	0.31	-0.3	-0.23	-0.25
Inhib_Constant	25.43	36.4	84.54	80.98
Intermol_energy	-7.76	-7.55	-7.64	-7.67
Vdw_hb_disolve_energy	-7.75	-7.54	-7.56	-7.66
Electrostatic energy	-0.01	-0.01	-0.09	-0.01
Total_internal	-0.41	-0.39	-0.81	-1.91
Torsional energy	1.49	1.49	2.09	2.09
Unbound energy	-0.41	-0.39	-0.81	-1.91
refRMS	53.02	50.32	54.5	51.18

Azine 1 demonstrated the lowest binding affinity of -6.27 kcal/mol and exhibited several interactions with the breast cancer protein 2IOK. These interactions included one π -cation interaction with the amino acid ARG 394, with a bond length of 4.13 Å, one π -anion interaction with GLU 353, with a bond length of 3.58 Å, two π - σ interactions with ILE 326, with bond lengths of 3.93 Å and 4.00 Å respectively, one π -anion interaction with GLU 323, with a bond length of 3.96 Å, and one π -alkyl interaction with the amino acid LEU 320, with a bond length of 4.85 Å and displays twelve hydrophobic interactions with PHE 404(A),

ARG 394(A), ILE 326(A), TRP 393 (A), GLU 323(A), LEU 320(A), GLY 442(A), VAL 446(A), PHE 445(A), PRO 325(A), LEU 327(A) and GLU 353(A) (Figure 2).

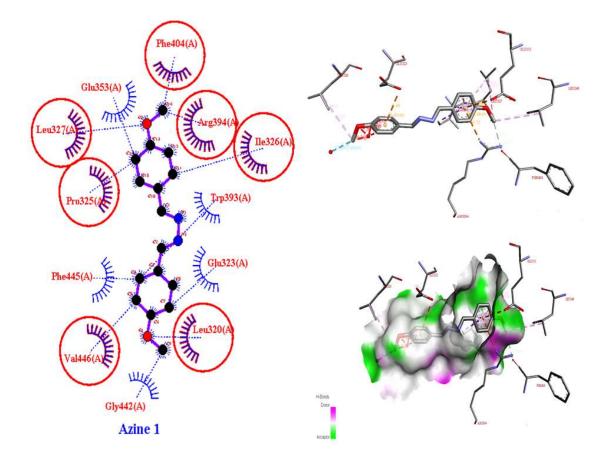


Figure 2. Docking image of azine 1 against 2IOK

Azine 2, with a binding affinity of -6.06 kcal/mol, exhibited several interactions with amino acids. Specifically, it formed one π -cation interaction with ARG 394, a π -alkyl interaction with PRO 324, a π -sulfur interaction with ARG 394, and two π - π T-shaped interactions with ARG 394, with bond lengths of 4.74Å, 4.68Å, 5.35Å, 4.73Å, and 4.91Å respectively. Additionally, the hydrogen atom of Azine 2 formed a hydrogen bond with the nitrogen atom of the amino acid TRP 393(A). Furthermore, Azine 2 exhibited ten hydrophobic interactions with GLU 443(A), GLY 442(A), GLU 323(A), PHE 445(A), ILE 386(A), GLY 390(A), LEU 387(A), LYS 449(A), MET 357(A), and GLU 353(A) (Figure-3).

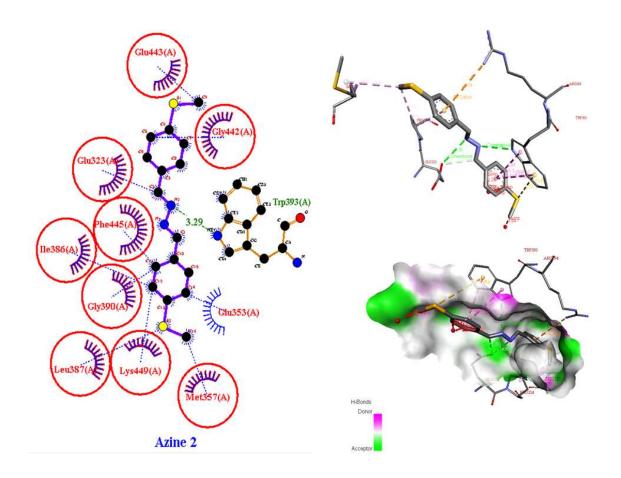


Figure 3. Docking image of azine 2 against 2IOK

The binding affinity of azine 3 with 2IOK is -5.56 kcal/mol, it having one hydrogen bond with LEU 1327(B), two π - π stacked, one π - σ , one π -cation and one π -alkyl interactions with amino acids TRP 1393, ILE 1326, ARG 1394 and PRO 1324 respectively (Figure 4).

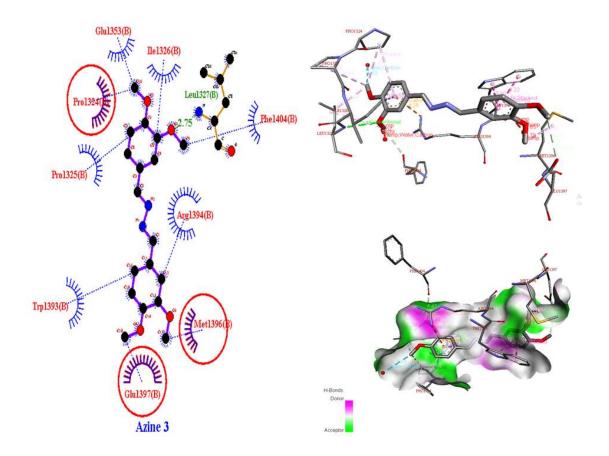


Figure 4. Docking image of azine 3 against 2IOK

Azine 4 demonstrated the binding affinity of -5.58 kcal/mol and exhibited several interactions with the breast cancer protein 2IOK. These interactions included one π -cation interaction with the amino acid ARG 1394, with a bond length of 3.52 Å, one π -anion interaction with GLU 1323, with a bond length of 3.69 Å, two π -alkyl interactions with PRO 1406, with bond lengths of 5.04 Å and 5.10 Å respectively, and displays twelve hydrophobic interactions and two hydrogen bond with PHE 404(A), ARG 394(A), ILE 326(A), TRP 393 (A), GLU 323(A), LEU 320(A), GLY 442(A), VAL 446(A), PHE 445(A), PRO 325(A), LEU 327(A), GLU 353(A), LEUL 1327(B) respectively (Figure 5).

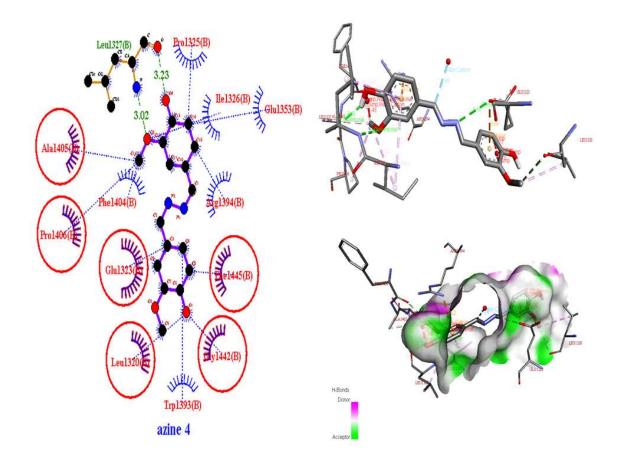


Figure 5. Docking image of azine 4 against 2IOK

In silico toxicological assessment

ProTox-II was used to calculate the compound's toxicity profile. Table 2 shows the predicted toxicity profile and Figure 6 display the Graphical representation of predicted dose value distribution for compound.

	Azin		Azine 1 Azine 2			Azine 3			Azine 4	
Classification	Target	Prediction	Probability	Prediction	Probability	Prediction	Probability	Prediction	Probabilit	
Organ toxicity	Hepatotoxicity	Inactive	0.72	Inactive	0.55	Active	0.54	Active	0.51	
Organ toxicity	Neurotoxicity	Active	0.72	Active	0.58	Inactive	0.60	Inactive	0.69	
Organ toxicity	Nephrotoxicity	Inactive	0.65	Inactive	0.81	Inactive	0.59	Inactive	0.51	
Organ toxicity	Respiratory toxicity	Inactive	0.81	Inactive	0.52	Inactive	0.61	Inactive	0.57	
Organ toxicity	Cardiotoxicity	Inactive	0.73	Inactive	0.72	Inactive	0.55	Active	0.50	
Toxicity end	Carcinogenicity	Inactive	0.55	Active	0.59	Active	0.52	Active	0.57	
points										
Toxicity end	Immunotoxicity	Inactive	0.58	Inactive	0.98	Inactive	0.58	Active	0.62	
points										
Toxicity end	Mutagenicity	Inactive	0.84	Inactive	0.55	Inactive	0.53	Inactive	0.56	
points										
Toxicity end	Cytotoxicity	Inactive	0.91	Inactive	0.84	Inactive	0.63	Inactive	0.79	
points										
Toxicity end	BBB-barrier	Active	0.70	Active	0.95	Active	0.83	Active	0.64	
points										
Toxicity end	Ecotoxicity	Active	0.74	Active	0.80	Active	0.66	Inactive	0.54	
points										
Toxicity end	Clinical toxicity	Inactive	0.70	Inactive	0.70	Inactive	0.68	Inactive	0.60	
points										
Toxicity end	Nutritional toxicity	Inactive	0.93	Inactive	0.68	Inactive	0.79	Inactive	0.77	
points										
Tox21-Nuclear	Aryl hydrocarbon	Inactive	0.86	Inactive	0.70	Active	0.56	Active	0.65	
receptor	Receptor (AhR)									
signalling										
pathways										
Tox21-Nuclear	Androgen Receptor	Inactive	0.98	Inactive	0.93	Inactive	0.93	Inactive	0.93	
receptor	(AR)									
signalling										
pathways										
Tox21-Nuclear	Androgen Receptor	Inactive	0.99	Inactive	0.98	Inactive	0.99	Inactive	0.99	
receptor	Ligand Binding									
signalling	Domain (AR-LBD)									
pathways										

Table 2. Complete Toxicity profile of azine 1-4

Tox21-Nuclear	Aromatase	Inactive	0.89	Inactive	0.92	Inactive	0.83	Inactive	0.89
receptor									
signalling									
pathways									
Tox21-Nuclear	Estrogen Receptor	Active	0.61	Inactive	0.71	Inactive	0.70	Inactive	0.62
receptor	Alpha (ER)								
signalling									
pathways									
Tox21-Nuclear	Estrogen Receptor	Inactive	0.89	Inactive	0.93	Inactive	0.88	Inactive	0.61
receptor	Ligand Binding								
signalling	Domain (ER-LBD)								
pathways									
Tox21-Nuclear	Peroxisome	Inactive	0.97	Inactive	0.95	Inactive	0.97	Inactive	0.92
receptor	Proliferator								
signalling	Activated Receptor								
pathways	Gamma (PPAR-								
	Gamma)								
Tox21-Stress	Nuclear factor	Inactive	0.85	Inactive	0.63	Inactive	0.85	Inactive	0.74
response	(erythroid-derived								
pathways	2)-like 2/antioxidant								
	responsive element								
	(nrf2/ARE)								
Tox21-Stress	Heat shock factor	Inactive	0.85	Inactive	0.63	Inactive	0.85	Inactive	0.74
response	response element								
pathways	(HSE)								
Tox21-Stress	Mitochondrial	Inactive	0.59	Inactive	0.78	Inactive	0.50	Active	0.64
response	Membrane Potential								
pathways	(MMP)								
Tox21-Stress	Phosphoprotein	Inactive	0.89	Inactive	0.94	Inactive	0.90	Inactive	0.79
response	(Tumor Supressor)								
pathways	p53								
Tox21-Stress	ATPase family	Inactive	0.84	Inactive	0.84	Inactive	0.85	Inactive	0.82
response	AAA domain-								
pathways	containing protein 5								
	(ATAD5)								
Molecular	Thyroid hormone	Inactive	0.90	Inactive	0.90	Inactive	0.90	Inactive	0.90
Initiating Events	receptor alpha								

(THRa)								
Thyroid hormone	Inactive	0.78	Inactive	0.78	Inactive	0.78	Inactive	0.78
receptor beta								
(THRβ)								
Transtyretrin (TTR)	Inactive	0.97	Inactive	0.97	Inactive	0.97	Inactive	0.97
D II (T.	0.00	.	0.00	T	0.00		0.00
	Inactive	0.98	Inactive	0.98	Inactive	0.98	Inactive	0.98
(RYR)								
GABA receptor	Inactive	0.96	Inactive	0.96	Inactive	0.96	Inactive	0.96
(GABAR)								
Glutamate N-	Inactive	0.92	Inactive	0.92	Inactive	0.92	Inactive	0.92
methyl-D-aspartate								
receptor (NMDAR)								
alpha-amino-3-	Inactive	0.97	Inactive	0.97	Inactive	0.97	Inactive	0.97
hydroxy-5-methyl-								
4-								
isoxazolepropionate								
· · · /	T	0.00	Turnet	0.00	Lucia	0.00	Turting	0.00
1	Inactive	0.99	Inactive	0.99	Inactive	0.99	Inactive	0.99
Achetylcholinestera	Inactive	0.72	Inactive	0.55	Active	0.54	Active	0.51
se (AChE)								
Constitutive	Inactive	0.98	Inactive	0.98	Inactive	0.98	Inactive	0.98
androstane receptor								
(CAR)								
Pregnane X receptor	Inactive	0.92	Inactive	0.92	Inactive	0.92	Inactive	0.92
` ´	Inactivo	0.07	Inactivo	0.07	Inactiva	0.07	Inactiva	0.97
	macuve	0.97	mactive	0.97	mactive	0.97	mactive	0.9/
Voltage gated	Inactive	0.95	Inactive	0.95	Inactive	0.95	Inactive	0.95
sodium channel								
(VGSC)								
Na+/I- symporter	Inactive	0.98	Inactive	0.98	Inactive	0.98	Inactive	0.98
(NIS)								
Cytochrome	Active	0.68	Inactive	0.51	Active	0.66	Inactive	0.55
		1	1		1			
	Thyroid hormone receptor beta (THRβ) beta Transtyretrin (TTR) Ryanodine receptor (RYR) receptor GABA receptor (GABAR) receptor Glutamate N- Glutamate N- GabbaAR receptor isoxazoleprosonate receptor fachatyl=D-aspartate receptor alpha-amino-3- etal isoxazoleprosonate receptor facatol receptor Kainate receptor receptor (ACAR) use KAchetyl=D-insetera receptor isocazole receptor use GAChetyl=D-insetera receptor (CAR) use isocacione receptor (CAR) use (Parganae receptor (PAR) use (NADH-quineus use (NADHONE use (VOSC) use (Na+/I) symporter (NAS) <t< td=""><td>ThyroidhormoneInactivereceptorbeta(Inactive(THRβ)InactiveTranstyretrin (TTR)InactiveRyanodineInactive(RYR)Inactive(GABAreceptorInactive(GABAR)Inactive(GABAR)Inactivereceptor (NMDAR)Inactivealpha-amino-3-Inactivehydroxy-5-methyl-InactiveisoxazolepropionateInactivereceptor (AMPAR)InactiveKainatereceptorInactiveInactivekainatereceptorKAR)InactiveisoxazolepropionateInactivereceptor 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Metabolism	Cytochrome	Active	0.67	Inactive	0.65	Inactive	0.54	Active	0.59
	CYP2C19								
Metabolism	Cytochrome CYP2C9	Active	0.81	Inactive	0.54	Active	0.52	Inactive	0.50
Metabolism	Cytochrome CYP2D6	Inactive	0.61	Inactive	0.55	Inactive	0.64	Inactive	0.69
Metabolism	Cytochrome CYP3A4	Inactive	0.77	Inactive	0.68	Active	0.51	Inactive	0.51
Metabolism	Cytochrome CYP2E1	Inactive	0.99	Inactive	0.79	Inactive	0.99	Inactive	0.99

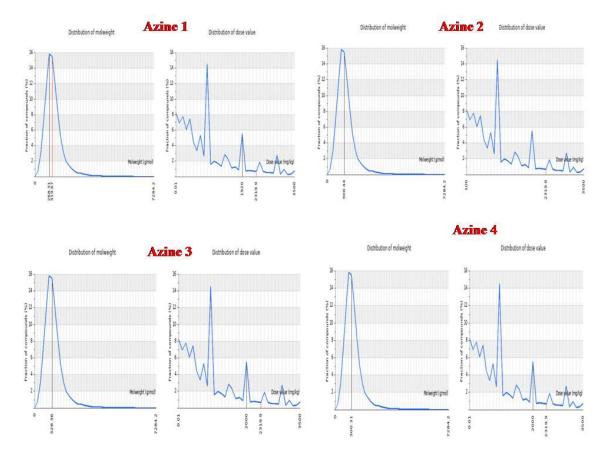


Figure 6. Graphical representation of predicted dose value distribution for azine 1-4

Conclusion

In conclusion, our study focused on the docking of four symmetrical azines, previously reported by our team, with the breast cancer protein 2IOK. Among these azines, azine one displayed notably higher binding affinity than the others, indicating its potential as a lead compound for further development. Furthermore, the toxicity of all four azines was assessed using ProTox-II software, providing valuable insights into their safety profiles. These findings contribute to the understanding of the molecular interactions of these azines with the target protein and their potential therapeutic relevance in breast cancer treatment.

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