

## Article

# Potential Hydrophobic Pocket of Squalene Synthase: An In Silico Analysis

### Article Info

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**Abstract.** Cardiovascular disease cases increase due to consumption cholesterol dietary habit. It is well-known that squalene synthase (SQS) is the first committed enzyme for cholesterol synthesis. Therefore, SQS become target of anti-cholesterol. This paper aims to determine the potential binding pocket of SQS (PDB ID: 1EZP). Dogsitescorer, siteFinder, and DEPTH were used for binding pocket prediction and MOE 2009.10 was performed for molecular docking. We found that there are five out of 37 pockets which have druggability score above 0.8. Pocket\_5 is the highest drugability and favorable for hydrophobic interaction, yet lower number of hydrogen bond with the ligand. However, Pocket\_2, and Pocket\_3 are suitable for hydrogen bond formation of ligand-protein. Molecular docking study showed that TAK-475, D99, and Cynarin inhibitors were embedded on the P\_2 and P\_3 of SQS, showing that P2 and P3 are promising binding pocket for ligand interactions. These results show a promising alternative to design anti-cholesterol using these potential pocket in silico.

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## 1. Introduction

High levels of cholesterol in dietary increases the risks of heart disease and stroke [1]. It has been reported by WHO that around 17 million people die by heart disease and stroke [2]. The prevalence of cholesterol cases increasing is closely related to the country's economy. WHO reports that cholesterol cases in high-income countries increase more than 50% of adults, whereas, in low-income countries, it is only 25% of adults [2].

The current hypercholesterol treatments include low cholesterol diet, exercise, inhibitors of cholesterol adsorbtion, and inhibitors of cholesterol synthesis [3]. The most advanced anti-cholesterol drug against HMG-CoA reductase, statin family, showed undesirable side effects. Statins produce statin-induced myotoxicity (SIM) manifest with severe muscle weakness, muscle pain, and muscle tenderness [4]. In addition, HMG-CoA reductase inhibition prevents mevalonate synthesis which is imporant precursor for dolichol, ubiquinone, and RAS [5].

Currently, a promising strategy for blocking cholesterol synthesis is the development of inhibitor for Squalene synthase (SQS). SQS plays an important role as an enzyme for cholesterol biosynthesis which catalyzes the first step of the steroid synthesis pathway [6]. Clinical studies have shown that inhibition of SQS are effective in lowering serum cholesterol concentrations and LDL-C without interrupting isoprenoid production [7][8]. Moreover, SQS inhibitors have have fewer secondary effects than HMG-CoA reductase inhibitors[9] Due to its main role and strategic location in cholesterol synthesis, SQS is potential drug target for the treatment of cholesterol disease.

In order to design the new SQS inhibitors, in silico strategies is considered as the first step of drug development strategy. Huang et.al (2019) reported that three out of 373,782 inhibitors were identified as potential inhibitors of SQS using in silico screening and in vitro studies[6]. Other studies on traditional chinese medicine showed that cynarin, D99, and TAK-475 inhibitors had strong binding in the dynamic system with SQS protein (PDB ID: 3ASX) [7]. Most of these docking studies utilized blind docking. Blind docking, a method for ligand-protein binding prediction, can be used for mapping of drug development. A key advantage of blind docking is the capability to predict the binding without any prior knowledge of the target pocket [10]. However, limitations of blind docking are the unknown number of trials and energy evaluations. It is recommended that more than 100 times of trial, and 10 million energy evaluations [11]. Due to these limitation, therefore, in the in silico drug discovery, the identification of ligand-protein binding pocket is the first stage, followed by hit identification, lead optimization, and ADMET properties calculation [12][13]. Recently, docking protocol using known binding pocket increases accuracy and efficiency than blind docking only with the lower time of trial [14]. Thus, the present study aimed to determine the most potential pocket of squalene synthase for cholesterol synthesis inhibition.

## **2. Method**

### **2.1. Hardware, Software and Webserver**

Personal laptop HP model 14-ck0011TU Intel® Celeron® N4000 CPU 1.1GHz, RAM 4.0 GB, Microsoft Windows 10 Pro 64-bit with internet connection. MOE 2009.10 software (Chemical Computing Group ULC). Webserver <http://www.rcsb.org/pdb/>, <https://proteins.plus/>, and <http://cospi.iiserpune.ac.in/depth>.

### **2.2. Material**

Human squalene synthase (PDB ID: 1EZF) was obtained from the Protein Data Bank (PDB) database (<http://www.rcsb.org/pdb/>). Ligands structure of TAK-475, D99, and cynarine were retrieved from Pubchem database (<https://pubchem.ncbi.nlm.nih.gov/>) with the CID : 9874248, 54669582, and 5281769, respectively.

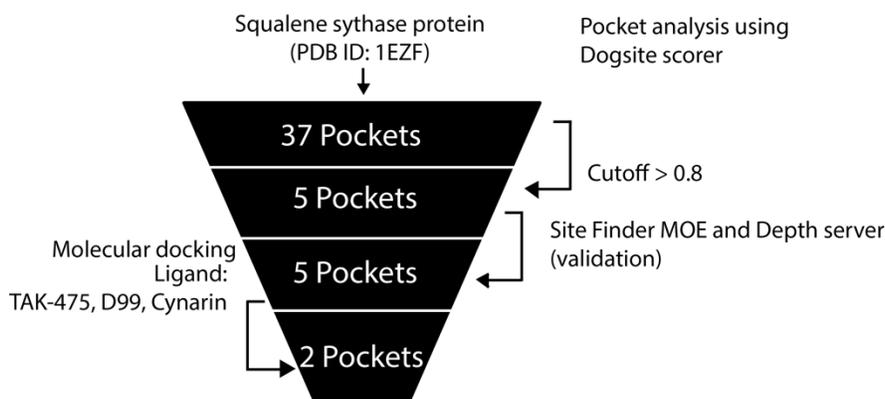
### **2.3. SQS Binding Pocket Analysis**

To predict the binding pocket of human squalene synthase, we have performed the binding pocket analysis. Dogsite scorer server (<https://proteins.plus/>) [15], Site Finder MOE2009.10 software

[16], and Depth server (<http://cospi.iiserpune.ac.in/depth>) [17] were used to measure the volume, druggability, and polarity of the pockets.

#### 2.4. Molecular Docking and Binding Analysis

To further identify the potential binding pocket for inhibitors of human squalene synthase, we have docked 1EZf with TAK-475, D99, and cynarine. Ligands were optimized using MOE 2009.10 [18]. Preparation of Human squalene synthase was carried out using MOE 2009.10 software. The native ligands and water molecules were separated from the human SQS structure (PDB ID: 1EZf) and optimized using current forcefield, adjust hydrogen and lone pairs, gradient 0,05, and forcefield partial charges calculation [19]. The docking study was performed by MOE 2009.10 software, which uses; triangle matcher with 2500000 iterations as placement, one-time rescoring, London dG, 100 repetitions for the first retain, and force field refinement [18] [19].



**Figure 1.** The protocol Utilized in this study for discovering potential pockets

### 3. Results and Discussion

Binding pocket determination is important to increase accuracy and efficiency in molecular docking. In this study, we analysed the druggability, volume, and polarity of pockets, and ligand-protein interaction to predict the potential binding pocket. SQS structure (PDB ID: 1EZf) was utilized to analyse the druggability, volume, and polarity of pockets. The cutoff of 0.7 was used for druggability score as the based on the average score [20].

Dogsitescorer result showed that there are seven pockets out of fifteen which have the druggable score  $> 0.70$  and also have a high number of volume (Table 1), indicating discerning between the druggable and non-druggable pockets. The pocket volume increase with the increase of druggability (Table 1). Volume is important variable for druggable prediction, which is individually calculated using the number of volume grid points and grid spacing [21]. However, pocket 5 (P\_5) having low pocket volume reached the highest druggability score. It might be due to the percentage of hydrophobic (nonpolar) area is higher than hydrophilic area, but not too different. A similar phenomenon in which SARS-Cov-2 S-glycoprotein binding pocket prediction has been reported previously [22]. Previous report highlighted that water molecules trapped inside the binding site is crucial for hydrophobicity and hydrophilicity of the binding site. Therefore, maintaining the water balance is important to ligand recognition.

**Table 1.** Potential binding pockets of Squalene synthase (PDB ID: 1EZF) based on Drugable score analysed by Dogscorer server

Pocket <sup>a</sup>	Drugable score	Volume (Å <sup>3</sup> )	Nonpolar (%)	Polar (%)	Charge (%)
P_5	0.84	536.59	46	40	14
P_0	0.83	768.73	48	22	30
P_1	0.83	731.62	45	24	31
P_2	0.81	609.07	64	28	8
P_3	0.80	560.22	58	35	8
P_4	0.79	541.14	60	32	8
P_6	0.75	523.64	37	19	45
P_7	0.66	306.72	35	40	25
P_12	0.63	211.83	62	31	8
P_11	0.62	237.40	32	42	26
P_9	0.60	278.89	23	69	8
P_10	0.60	266.81	69	31	0
P_8	0,59	298.49	65	29	6
P_13	0.54	211.48	33	47	20
P_14	0.43	206.58	33	33	34

<sup>a</sup>Pocket 1, Pocket 2, etc are shown as P\_1, P\_2, etc

In order to provide better discerning between the druggable and non-druggable pockets, we next utilized the pockets which have score  $\geq 0.8$ . Binding pocket analysis were confirmed using sitefinder MOE, and Depth. Table 2 showed that all the pockets have more hydrophobic residues rather than hydrophilic. Unlike P\_0, P\_1 and P\_5, dogscorer and sitefinder analysis result on P\_2 and P\_3 were similar. Previous study revealed that the similarity results of the dogscore and sitefinder analysis result in better complex ligand and protein interaction [22].

To confirm the capability of pockets as binding pocket, we docked all five pocket with TAK-475, D99, dan cynarin ligand. H-bond residues for SQS were provided in Tables 3. We showed that there are more hydrogen bonds than hydrophobic interaction. Consistently, Ligand-OppA protein interaction was achieved by utilizing the hydrogen bonds and the electrostatic forces[23]. Moreover, other findings revealed that hydrogen bonds are considered as the main facilitators of protein-ligand interaction [24]. The present study showed that TAK-475 has the highest number of hydrogen bonds in P\_3, while cynarin has high number of hydrogen bonds in P\_1 and P\_5. Taken together, these results suggest that P\_2 and P\_3 are potential as binding pocket. As the ligands are completely embedded within the pocket protein Figure 1, physicochemical properties of ligand are important to consider ligand-protein interaction. Both TAK-475 and cynarin have high number of hydrogen acceptors and donors, suggesting strong protein-ligand interactions due to physicochemical properties (Table 4). Our result is supported by a recent study showing that hydrogen acceptor and donor of the ligands are the main facilitators of ligand-protein interaction[22]. Therefore, physicochemical properties of ligand and pocket hydrophobicities are suggested as the main factor for the drug design of cholesterol disease.

**Table 2.** Residues of Amino Acid in binding pockets of SQS

Pocket <sup>a</sup>	Hydrophobicity		Charges		Special charges	
	Hydrophobic	Hydrophilic	(+) es	(-) es		
P_0	MET_B154 <sup>b</sup>	ILE_B217 <sup>c</sup>	GLN_B212	LYS_B160	GLU_B83	PRO_B232
	PHE_B157	TYR_B220	ASN_B215	HIS_B161	ASP_B84	
	LEU_B158	PHE_B230 <sup>d</sup>	GLN_B224	ARG_B22	ASP_B219	
	VAL_B162	TRP_B231	GLN_B233	8	ASP_B223	
	TYR_B171	TRP_B236			GLU_B22	
	VAL_B175	LEU_B243			9	
	ILE_B216					
P_1	MET_A86	VAL_A175	GLN_A212	LYS_A160	ASP_A80	GLY_A227
	MET_A150	ILE_A216	ASN_A215	HIS_A16	GLU_A83	PRO_A232
	MET_A154	TYR_A220	GLN_A224	1	ASP_A84	
	PHE_A157		GLN_A233	ARG_A2	ASP_A21	
	LEU_A158	PHE_A230		28	9	
	VAL_A162	TRP_A231			ASP_A22	
	TYR_A171	TRP_A236			3	
	LEU_A243			GLU_A22		
				9		
P_2	PHE_C54	VAL_C179	SER_C184	ARG_C77	ASP_C80	TYR_C180GLY_C
	ILE_C58	LEU_C183	GLN_C293			C208CYS_C
	VAL_C69	PHE_C187				289PRO_C2
	PHE_C72	ALA_C204				92
	TYR_C73	MET_C207				
	LEU_C76	LEU_C211				
	MET_C150	TYR_C276				
VAL_C175	PHE_C288					
ALA_C176						
P_3	PHE_A54	VAL_A179	SER_A184	ARG_A77	ASP_A80	TYR_A177GLY_A
	ILE_A58	LEU_A183	GLN_A212			A180GLY_A
	VAL_A69	PHE_A187	GLN_A293			A208CYS_A
	PHE_A72	ALA_A204				289PRO_A2
	TYR_A73	MET_A207				92
	LEU_A76	LEU_A211				
	VAL_A175	TYR_A276				
ALA_A176	PHE_A288					
P_5	PHE_A54	ILE_A291	ASN_A48	ARG_A52	GLU_B343	PRO_B332,
	ALA_A55	ALA_B333	GLN_A49	ARG_B36		
	ALA_A56	ALA_B336	SER_A51	7		
	VAL_A57	ILE_B337	SER_A53	HIS_B347		
	ILE_A58	TYR_B339	GLN_A283			
	GLN_A59	MET_B342	SER_A284			
	ALA_A60	TYR_B346	ASN_A287			
	LEU_A61	ILE_B363	THR_A329			
	PHE_A288	ILE_B366	ASN_B330			
		GLN_B340				
		SER_B364				

<sup>a</sup>Pocket 1, Pocket 2, etc are shown as P\_1, P\_2, etc

<sup>b</sup>residues with red colour = predicted by dogsitescorer and sitefinder moe

<sup>c</sup>residues with black colour = predicted by dogsitescorer only

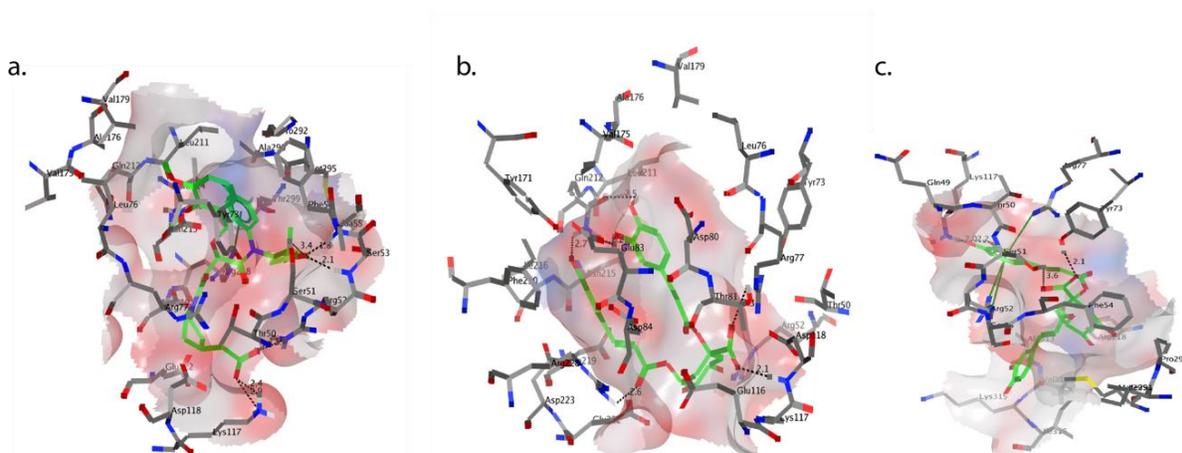
<sup>d</sup>residues with yellow highlight = predicted by dogsitescorer, sitefinder moe, and DEPTH server

**Table 3.** Binding energy and residues involved in ligands interaction

Pocket	Ligand	Binding energy	Residues involved	
			Hydrogen bond (distance Å <sup>0</sup> )	Hydrophobic interaction (distance Å <sup>0</sup> )
P_0	TAK-475	-17.1669	TYR_B171 (3.37) ARG_B228 (1.99)	ARG_B77 (3.00)
	D99	-15.5290	TYR_B171 (3.06) ARG_B228 (2.05)	
	Cynarin	-16.1079	GLU_B116 (1.99) ASP_B219 (1.99)	PHE_B230
P_1	TAK-475	-15.0187	ARG_A52 (2.02) ARG_A77 (2.04) LYS_A117 (2.57) GLN_A212 (1.84)	-
	D99	-12.6647	ARG_A77 (2.03) ASP_A219 (1.67)	PHE_A230
	Cynarin	-16.2647	THR_A81 (3.31) ASP_A118 (2.14) ARG_A228 (2.57) GLN_A212 (2.71)(2.54) GLU_A83 (2.23)	-
P_2	TAK-475	-17.3370	ARG_C52 (1.78) LYS_C117 (2.1)	-
	D99	-15.8893	ASN_C215 (2.71) LYS_C117 (1.97)	ARG_C52 (2.66)(3.16)
	Cynarin	-16.3325	ARG_C52 (1.99) VAL_C69 (1.97)	-
P_3	TAK-475	-14.8776	ASN_A215 (2.59) SER_A51 (3.41) PHE_A54 (1.83) SER_A53 (2.05) ARG_A52 (1.92) LYS_A117 (2.43)	ARG_A218
	D99	-13.5409	ARG_A52 (1.93) LYS_A117 (2.2)	-
	Cynarin	-16.0648	SER_A53 (2.29) TYR_A73 (3.46) ARG_A77 (2.38)	-
P_5	TAK-475	-12.3923	SER_A53 (3.35)	ARG_A52 (3.01)
	D99	-12.7368	ARG_A52 (1.89) ARG_A218 (2.29) LYS_A117 (2.32)	ARG_A218
	Cynarin	-14.7047	LYS_A117 (4.98)(2.19) SER_A51 (3.56) TYR_A73 (2.13)	ARG_A52, ARG_A77

**Table 4.** Physiochemical properties of ligand

No	Ligan	Properties		
		MW	H Acceptor	H Donor
1	TAK-475	645.1	9	1
2	D99	549.5	5	2
3	Cynarin	516.4	12	7



**Figure 2.** SQS Local Structure of the Docking Complexes: a) TAK-475 with P\_3, Cynarin with b) P\_1 and c) P\_5

#### 4. Conclusion

In conclusion, we have identified two SQS hydrophobic pocket (P\_2, and P\_3) as potential binding pocket. The findings from molecular docking demonstrated that TAK-475, D99, and Cynarin inhibitors are embedded on the P\_2 and P\_3 of SQS. Complex of inhibitors and P\_3 has lower binding energy and higher hydrogen bond than that and P\_2 of SQS. These findings provide promising alternative to design anti-cholesterol.

#### 5. Acknowledgment

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