Structure-based Discovery of Potential Inhibitors Targeting Sodium-bile Acid Co-transporter of Carcinogenic Liver Fluke Clonorchis Sinensis

Won Gi Yoo
Chung-Ang University

Fuhong Dai
Soochow University

Jhang Ho Pak
University of Ulsan College of Medicine

Sung-Jong Hong (✉ hongsj@cau.ac.kr)

Research

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Abstract

**Background:** *Clonorchis sinensis* requires bile acid transporters as this fluke inhabits bile juice-filled biliary ducts, which provide an extreme environment. *C. sinensis* sodium-bile acid co-transporter (CsSBAT) is indispensable for survival in the final host, as it circulates taurocholate and prevents bile toxicity in the fluke; hence, it is recognized as a useful drug target.

**Results:** In the present study, using structure-based drug discovery approach, we presented inhibitor candidates targeting a bile acid-binding pocket of CsSBAT. CsSBAT models were built using comparative tertiary structure modeling based on a bile acid transporter template (PDB ID: 3zuy and 4n7x), and were applied into AutoDock Vina for competitive docking simulation. First, potential compounds were identified from PubChem (holding more than 100,000 compounds) by applying three criteria: i) interacting more favorably with CsSBAT than with a human homolog, ii) intimate interaction to the inward- and outward-facing conformational states, iii) binding with CsSBAT more preferably than natural bile acids. Second, two compounds were identified following Lipinski’s rule of five. Third, another two compounds of molecular weight higher than 500 Da (Mr > 500 Da) were presumed to efficiently block the transporter via a feasible rational screening strategy. These four inhibitor candidates exhibited least toxicity that may enhance drug-likeness properties.

**Conclusion:** It is proposed that four compounds act as potential inhibitors toward CsSBAT, and further studies are warranted for drug development process against clonorchiasis.

Background

*Clonorchis sinensis* is a trematode parasite commonly observed in humans, and is transmitted by eating raw or undercooked freshwater fish contaminated with its metacercariae [1]. The intensity and duration of the infection determine the disease complications caused by this trematode. World Health Organization has classified *C. sinensis* as a Group 1 biological carcinogen inducing cholangiocarcinoma in humans [2].

Praziquantel has been strongly recommended for treating trematode infections in humans including clonorchiasis [3]. Due to the extensive use of praziquantel, certain trematodes in the tropical countries have developed low sensitivity to this drug [4, 5]. Hence, new chemotherapeutic agents should be developed to circumvent the low sensitivity or drug resistance in these trematodes. To date, several drug candidates for trematodiasis have been tested on animals [6–10]. Among these, numerous drugs are under trials for clonorchiasis patients [11].

Structure-based drug discovery (SBDD) is a powerful approach for compound identification, and has recently gained immense attention in drug development [12]. Several studies have reported the effectiveness of computational approaches in discovering novel antiparasitic druggable compounds [13–15]. For *C. sinensis*, secretory phospholipase A2 enzyme [16] and 20.6-kDa tegumental protein [17] were computationally evaluated as druggable targets; however, more compounds designed by SBDD are
necessary for expanding drug candidate repertoire because they can be optionally tested according to the worm burden or developmental stages of *C. sinensis*.

Notably, bile acids trigger physiological stimuli; however, these metabolites can have detrimental effects on *C. sinensis* survival, as evidence suggests that the accumulated bile acids can cause toxicity to the worm's tissues and cells [18, 19]. Recent studies reported that bile components can be toxic or detrimental to *C. sinensis* survival [10, 20, 21]. Under *in vitro* conditions, higher concentration (> 0.005%) of bile and lithocholic acid significantly decreased the worm survival [10, 20, 21]. Therefore, a defense system against the accumulation and toxicity of bile acids could be targeted to decrease *C. sinensis* survival.

Apical sodium-dependent bile acid transporter (ASBT) and Na\(^+\)-taurocholate co-transporting polypeptide (NTCP) contribute to the enterohepatic circulation of bile salts in humans as homologs of *C. sinensis* sodium-bile acid cotransporter (CsSBAT) [22]. Since ASBT is a promising druggable target due to effectiveness in bile circulation, several inhibitors have been developed and evaluated [23–25]. Therefore, CsSBAT should be considered as a druggable target, whose taurocholate-binding pocket can be blocked and further inhibition can impede its transporter function.

In the present study, CsSBAT structures were reliably prepared using comparative tertiary structure modeling and refinements. For compound screening, the Lipinski's rule of five and the rational virtual screening strategy were applied. Here, we identified putative inhibitory compounds, which competitively targeted the taurocholate-binding pocket of CsSBAT.

**Methods**

**Tertiary structure modeling and refinement**

The full-length cDNA sequence of CsSBAT (Acc. No. KX756671) was retrieved from GenBank database [26]. To generate three-dimensional (3D) structure of CsSBAT, we compared the 3D structures built using different 3D modeling softwares such as Swiss-Model [27], IntFOLD [28], Phyre2 [29], RaptorX [30], HHpred [31], and I-TASSER [32]. The 3D models were then refined in two steps. First, low free-energy conformations of the 3D structure were refined by full-atomic simulations using either ModRefiner [33] or FG-MD [34]. Thereafter, backbone and side chains of the structure were refined using GalaxyRefine [35] using “both mild and aggressive relaxation” method based on repeated perturbation and overall conformational relaxation with short molecular dynamics simulations. Outward-facing (OF) and inward-facing (IF) conformations of CsSBAT were modeled using the templates *Yersinia frederiksenii* (YfASBT; PDB ID: 4n7x_A) [36] and *Neisseria meningitides* ASBT (NmASBT; PDB ID: 3zuy_A) [37], respectively.

**Quality validation and binding pockets of 3D models**

Potential errors in the 3D models were evaluated using PROCHECK [38], ProSA [39], and ERRAT [40]. Residue-by-residue stereochemical quality of 3D models was verified by Ramachandran plot [41] of PROCHECK. Overall quality score was analyzed by calculating atomic coordinates of the model using ProSA with a Z-score of experimentally determined structures deposited in PDB [42]. Statistics of
nonbonded atom–atom interactions were validated in comparison to a database of reliable high-resolution crystallographic structures using ERRAT \[40\]. All structures and protein–compound interactions were visualized using UCSF Chimera v1.14 \[43\]. Disordered region was predicted using ESPritz \[44\]. Substrate-binding sites in CsSBAT were predicted using COACH \[45\], referring to the analogs with similar binding sites.

**In silico screening of putative inhibitors**

Two strategies were applied to screen the putative inhibitors against IF-/OF-CsSBAT and IF-/OF-*Homo sapiens* ASBT (HsASBT) (Fig. 1): i) Compounds (99,288 ea) following the Lipinski’s rule of five \[46\] were screened against “diverse-lib” compounds database using MTiOpenScreen \[47\]. A list of compounds from each dataset was compared using Venn diagrams generated via jvenn \[48\]; ii) Compounds (1,255 ea) with high molecular weight (Mr) ranging 500–1,200 Da were collected from PubChem database \[49\] as of June 2018 and were screened using AutoDock Vina v1.1.2 \[50\]. Both docking tools were run with following parameters specifying substrate-binding pocket. Grid center parameters were set with X:77.63, Y:76.50, and Z:63.58 for IF conformation, and X:65.26, Y:74.20, and Z:68.12 for OF conformation. These were calculated using Autoligand v1.10 \[51, 52\]. Grid sizes were set at X:40, Y:40, and Z:40. SDF and PDBQT formats were converted into MOL2 format using Open Babel \[53\]. Each protein–ligand interaction map was generated using ViewDock method of UCSF Chimera v1.14 \[43\] and visualized using LigPlot+ v1.4.5 \[54\]. Massive data were analyzed iteratively and parsed using in-house Python scripts. For positive controls, bile acids (7 ea) including taurocholic acid were collected from PubChem database \[49\].

**Toxicity assessment**

Toxicity risk and oral toxicity (LD$_{50}$) were predicted using ProTox \[55\]. Higher LD$_{50}$ dose led to less toxic compound. Toxicity class ranging 4–6 indicates that the compound is safe.

**Results And Discussion**

**The best 3D model of CsSBAT was built from comparative homology modeling**

Both N- and C-terminal regions of CsSBAT were predicted to be hypothetical as well as disordered. In particular, disordered regions can result in long simulation time and may lead to errors in the structural clustering process \[56\]. Therefore, these regions were excluded from building 3D models (Additional file 1: Fig. S1). Functional region (residues 185–492) of CsSBAT matched well with experimentally characterized ASBTs such as IF-NmASBT (PDB ID: 3zuy_A) \[37\] and OF-YfASBT (PDB ID: 4n7x_A) \[36\].

To obtain reliable homologous models of parasites, combined approach of 3D modeling methods and refinement were employed \[17–19, 57–59\]. All predicted 3D models of IF-CsSBAT were evaluated using homology modeling programs such as Swiss-Model \[27\], IntFOLD \[28\], Phyre2 \[29\], RaptorX \[30\], and
HHpred [31], and threading-based modeling program such as I-TASSER [32] (Table 1). Swiss-Model, IntFOLD, and HHpred revealed values greater than 91.0% in the most favored region of Ramachandran plot [41]. Except for IntFOLD presenting erroneous ERRAT value, Swiss-Model, I-TASSER, and HHpred were evaluated further in terms of refinement because I-TASSER is efficient in building structure of unaligned regions by employing \textit{ab initio} modeling [60].

### Table 1

| 3D modeling program | Verification value/score |
|---------------------|--------------------------|
|                     | Ramachandran\(^1\) | ERRAT | ProSA | Cov\(^2\) |
| Swiss-Model         | 91.0% (0.8%) | 46.0 | -3.15 | 0.56 |
| I-TASSER            | 75.8% (1.9%) | 92.0 | -1.07 | 1.00 |
| IntFOLD             | 93.8% (0.2%) | n.a. \(^3\) | -1.81 | 1.00 |
| Phyre2              | 73.2% (5.1%) | 26.9 | -3.29 | 1.00 |
| RaptorX             | 89.3% (0.4%) | 41.9 | -2.87 | 1.00 |
| HHpred              | 91.7% (0%)   | 50.2 | -1.43 | 1.00 |

\(^1\) Ramachandran plot, % of most favored region (disallowed region)

\(^2\) Cov: Sequence coverage, ratio of predicted region relative to whole region

\(^3\) n.a.: not available

Homology modeling of membrane proteins is challenging when the target protein shares low sequence identity (approximately 20%) with a template [61]. This issue was circumvented by refining the initial models indicating poor quality [17, 18]. Therefore, the initial models of IF-CsSBAT were refined with ModRefiner, FG-MD, and GalaxyRefine (Table 2). After comprehensive evaluation, we applied Swiss-Model for reliable 3D modeling, and thereafter, FG-MD and GalaxyRefine were used for effective refinement. In fact, Swiss-Model has been a powerful tool for transporter modeling [62, 63]. Refining enhanced the structural quality of the final model compared to that of the initial model, particularly in terms of ERRAT values, which were increased from 46.0–97.7% (Tables 1 and 2, Additional files 1 and 2: Figs. S1 and S2); however, I-TASSER and HHpred could not overcome poor values either in most favored regions of Ramachandran plot or in unacceptable ERRAT plot. Moreover, 3D models of OF-CsSBAT, OF-HsASBT and IF-HsASBT were prepared as aforementioned.
Table 2
Comparison on refined structural evaluation

| 3D modeling program | Refinement program | Verified score/value produced by |
|---------------------|--------------------|---------------------------------|
|                     | ModRefiner | FG-MD | GalaxyRefine | Ramachandran¹ | ERRAT | ProSA | Cov² |
| Swiss-Model         |           |      |              | 93.3% (0.8%) | 89.7 | -3.84 | 0.6  |
| I-TASSER            |           |      |              | 92.5% (1.2%) | 97.7 | -3.73 | 0.6  |
| HHpred              |           |      |              | 86.8% (1.9%) | 89.0 | -1.07 | 1.0  |
|                     |           |      |              | 74.7% (1.9%) | 92.9 | -1.24 | 1.0  |
|                     |           |      |              | 94.1% (0%)   | 69.2 | -1.89 | 1.0  |
|                     |           |      |              | 76.0% (1.3%) | 81.5 | -1.89 | 1.0  |

¹ Ramachandran plot, % of most favored region (disallowed region)
² Cov: Sequence coverage, ratio of predicted region relative to whole region
³ Symbol “□”: corresponding software was applied.
⁴ n.a.: not available

Bile acid-binding cavity and sodium-binding sites

For the translocation of bile acids across the cell membrane, alternative conformational changes in IF and OF conformations of secondary active transporters were proposed [36, 64]. One of the two conformational states was observed only in a particular state because it is difficult to crystallize the structure under other states. Structural information of one state was applied to predict another conformation of transporters [65]. Here, we predicted the OF-CsSBAT and IF-CsSBAT models based on OF-YfASBT [36] and IF-NmASBT [37], respectively. In CsSBAT, the bile acid-binding pocket was presumed to be formed with 9 residues in an extracellular cavity with a volume of 908 Å³ in OF-CsSBAT (Fig. 2A) and with 11 residues in an intracellular cavity with a volume of 986 Å³ in IF-CsSBAT (Fig. 2B). Among the pocket forming residues, five residues (Phe₁₉₆, Phe₂₂₂, Ala₂₈₈, Ala₂₉₁, and Met₂₉₅) participated in both conformations.

ASBT is a symporter dependent on Na⁺ gradient that drives bile acid transport, which carries one bile acid with two Na⁺ ions [66]. Two Na⁺-binding sites of CsSBAT, Na1 and Na2, were predicted based on NmASBT (Fig. 3A). Na1 site comprised Ser₂₈₉, Asn₂₉₀, Ser₃₀₃, Thr₃₀₇, and Glu₄₄₁ residues; whereas, Na2 site comprised Gln₂₅₂, Glu₄₄₁, Thr₄₄₂, Ile₄₄₄, and Gln₄₄₅ residues (Fig. 3B and C). These two sites were highly conserved with ASBTs and NTCPs [36]. Furthermore, Glu₄₄₁ was predicted to participate in Na⁺-binding of the two sites, implying that it might have a functionally significant role (Fig. 3C). In NmASBT, mutation at Glu₂₆₀ (corresponding to Glu₄₄₁ of CsSBAT) markedly alters the taurocholate transporting activity [37].
Recently, a putative third Na\(^+\)-binding site was proposed, albeit rather speculative without experimental evidence [64]. The third Na\(^+\)-binding site was reported from similar transporters such as glutamate [67] and leucine [68] transporters. In CsSBAT, residues Ile\(_{280}\), Gly\(_{281}\), Ser\(_{283}\), and Gln\(_{445}\) were predicted to act as binding sites of the third Na\(^+\) ion, which were superposed with corresponding residues on NmASBT (Fig. 3D). Residue Gln\(_{445}\) could be a key residue carrying Na\(^+\)-ion from Na2 to Na3 site (Fig. 3). Mutation of Gln\(_{258}\) in YfASBT (corresponding to Gln\(_{445}\) in CsSBAT) was reported to significantly reduce the Na\(^+\)-binding capacity of Na2 and Na3 sites [36]. Therefore, Glu\(_{441}\) and Gln\(_{445}\) in CsSBAT might act as molecular arms and transport Na\(^+\) ion from one site to the next.

For accurate molecular docking of CsSBAT against compounds in the library, grid center and size were precisely specified in the extracellular and intracellular bile acid-binding pockets of OF-CsSBAT (Fig. 2A) and IF-CsSBAT (Fig. 2B), respectively, rather than sodium-binding sites (Fig. 3B–D), as described in “Putative inhibitors screening” section of Materials and Methods.

**Putative inhibitors targeting at bile acid-binding pocket**

Structure-based virtual screening was used to select putative inhibitors of CsSBAT, which satisfied the following criteria (Fig. 1). i) A compound should interact more favorably with CsSBAT than with HsASBT to ensure accurate targeting. ii) OF-ASBT conformation should be considered as a target although IF-ASBT binding with taurocholate was used as a template for virtual docking [25], because the ASBTs transfer bile acid inward via conformational change from OF- to IF-conformation [69]. iii) The compound to be identified as a competitive inhibitor of taurocholate should reveal higher affinity than natural bile acids, ranging from −6.2 to −9.0 kcal/mol [25]. Theoretically, the binding energies of several bile acids against IF-CsSBAT and OF-CsSBAT conformations ranged from −6.1 to −8.7 kcal/mol (Table 3) [19]; however, those of HsASBT (CsSBAT homolog) were −9.0 kcal/mol and −9.2 kcal/mol against natural bile acids and poly(acrylic acid)-tetraDOCA conjugate (PATD; lead compound blocking HsASBT), respectively [25]. Thus, a cut-off value was set at −9.2 kcal/mol since the present study aimed to explore the most probable inhibitor candidates binding to CsSBAT.
Table 3
Binding energy for bile acids OF-/IF-SBAT of *Clonorchis sinensis* and OF-/IF-ASBT of *Homo sapiens* using AutoDock vina

| Bile acids                        | PubChem ID | *C. sinensis* (kcal/mol) | *H. sapiens* (kcal/mol) |
|-----------------------------------|------------|--------------------------|-------------------------|
| Chenodeoxycholic acid             | 24875071   | −6.7                     | −7.6                    |
| Taurochenodeoxycholic acid        | 312642451  | −6.4                     | −7.7                    |
| Glycochenodeoxycholic acid        | 177011774  | −6.4                     | −7.6                    |
| Glycodeoxycholic acid             | 57309861   | n.a.¹                    | n.a.                   |
| Deoxycholic acid                  | 222528²    | −6.1                     | −7.7                    |
| Taurocholic acid                  | 828139     | −6.6                     | −8.7                    |
| Glycocholic acid                  | 177011773  | −6.8                     | −8.5                    |

¹ n.a.: not available
² PubChem compound ID but other bile acids are PubChem substance ID

**Compounds satisfying Lipinski’s rule of five**

Compounds following the Lipinski’s rule of five [46] were screened using MTiOpenScreen [47]. Of the top 1,000 scoring compounds under docking simulation against OF- and IF-conformations of CsSBAT and HsASBT, 19 compounds that could interact with only OF-CsSBAT or IF-CsSBAT were selected (Fig. 4 and Additional file 3: Table S1). Of these, two compounds met our strict criteria. Compound 49734421 formed a hydrogen bond with Ala$_{291}$ of IF-CsSBAT and Asn$_{446}$ of OF-CsSBAT. Compound 124948115 formed two hydrogen bonds with OF-CsSBAT but not with IF-CsSBAT (Table 4 and Fig. 5). Majority of the residues of these two compounds were involved in hydrophobic interaction with residues on CsSBAT, implying that these interactions might play a crucial role in compound–protein interactivity. It has been reported that aromatic moieties with high hydrophobicity can enable beneficial interactions with nonpolar residues in the binding pocket [70].
Table 4
Inhibitory compounds subjected to virtual screening using MTiOpenScreen and selected by applying Lipinski’s rule of five

| PubChem ID | Binding energy (kcal/mol) | nRot | HBA | HBD | LogP | Mr | TPSA | Hydrogen bond | Toxicity class (LD₅₀: mg/kg) |
|------------|--------------------------|------|-----|-----|------|----|------|---------------|-----------------------------|
|            |                          | OF   | IF  |     |      |    |      | OF            |                             |
| 49734421   | −10.9                    | 4    | 7   | 1   | 2.3  | 396.5 | 94.8 | 1             | 4                           | (1,077)                     |
| 124948115  | −10.3                    | 5    | 6   | 1   | 2.8  | 397.5 | 67.6 | 2             | 0                           | (1,500)                     |

Abbreviations: nRot, number of rotatable bonds; HBA, hydrogen bond acceptors; HBD, hydrogen bond donors; LogP, lipophilicity; Mr, molecular weight; TPSA, topological polar surface area.

Since Lipinski’s rule of five was introduced in 1997 [46], absorption or drug permeability is presumed to be more likely when there are less than 5 hydrogen bond donors, less than 10 hydrogen bond acceptors, a molecular weight with less than 500, and a calculated LogP smaller than 5. Recently, it was suggested that antiparasitic drugs should be exempted from this rule because several drug leads for infectious diseases do not follow Lipinski’s rule of five [46, 71]. Less stringent criteria may allow to identify more lead compounds for further assays. The suggestion motivated us to find out more effective strategy for antiparasitic drugs.

**Large compounds satisfying high affinities**

PATD was recently synthesized and evaluated as a potent inhibitor against ASBT [25]. Molecular weight of PATD molecule is larger than 500 Da because it has several polyacrylic acids and tetraDOCAs. Surprisingly, PATD is a hydrophobic substrate, which violates ideal molecular weight of the Lipinski’s rule of five. Nonetheless, it successfully inhibits ASBT by filling up the bile acid-binding cavity [46]. This finding motivated us to screen compounds with molecular weight higher than 500 Da, which are assumed to tightly dock CsSBAT.

Compounds of high molecular weight (500–1,200 Da) were retrieved from PubChem compound library [49] and screened using AutoDock Vina v1.1.2 [50]. Of the 1,255 compounds, 49 compounds satisfied the three given criteria. By strictly applying the third criterion (higher affinity than natural bile acids), we selected 25 compounds with high affinity for CsSBAT, but low affinity for HsASBT (Additional file 4: Table S2). Five compounds presented toxicity values of 4–6 with LD₅₀ ranging 500–5,000 mg/kg (Table 5). Eventually, two compounds 45375808 (Fig. 6A and B) and 9806452 (Fig. 6C and D) were selected as possible candidate inhibitors for CsSBAT because these compounds could form more than two hydrogen bonds each with OF-CsSBAT and IF-CsSBAT.
Notably, through docking simulation on compound–protein interactions, residues Glu$_{229}$ and Gly$_{287}$ participated in hydrogen bonding in taurocholate-IF-CsSBAT complex, whereas residues Gly$_{287}$, Gln$_{345}$, and Gln$_{348}$ participated in hydrogen bonding in taurocholate-OF-CsSBAT complex (Additional file 2: Fig. S2). Gly$_{287}$ was involved both in taurocholate–IF-CsSBAT and taurocholate–OF-CsSBAT complexes. In majority of the compound–OF-CsSBAT complexes, this conserved residue Gly$_{287}$ was involved in hydrogen bond interaction (Fig. 6A, C, E, and G), and Gln$_{345}$ was involved in compound 92727-OF-CsSBAT complex formation (Fig. 6I). Among compound-IF-CsSBAT complexes, only compound 441243 formed hydrogen bond with Glu$_{229}$ (Fig. 6F). Compared to taurocholate-CsSBAT complexes, Ala$_{288}$ acted as a key residue for either hydrogen bonds or hydrophobic interactions in three compound–CsSBAT complexes (Fig. 6A–F). Compounds 441243 and 3693566 presented 5 and 6 hydrogen bonds, respectively; however, these compounds were excluded owing to less hydrogen bonds in compound-IF-CsSBAT interactions, which could result in off-target binding with adverse drug reactions [72].

Compound 45375808, known as sofosbuvir, was proposed to inhibit nonstructural protein 5B polymerase in hepatitis C virus (HCV) [73]. Analogs of this compound exhibited inhibitory effect on HCV [74]. Compound 9806452 was reported as an inhibitor of matrix metalloproteinases [75] such as gelatinase A associated with tumor metastasis [76] and stromelysin-1 found in osteoarthritis [77]. Carboxyalkyl peptides
containing a biphenylylethyl group inhibits adult *Schistosoma mansoni* [78]. Considering these reports, it is suggested that compounds 45375808 and 9806452 could be anthelminthic candidates for *C. sinensis*.

**Conclusion**

*C. sinensis* sodium-bile acid cotransporter (CsSBAT) should be indispensable for survival in the bile duct. Inhibition of the bile transporters may perturb bile acid transport, and prove detrimental to *C. sinensis*. We identified the inhibitory compounds targeting the bile acid-binding pockets in CsSBAT, based on the physiological essence of crucial importance using structure-based drug discovery approach. First, PubChem compounds 49734421 and 124948115 were selected by applying the Lipinski’s rule of five. Furthermore, we devised a feasible rational screening strategy to search inhibitor candidates with molecular weight greater than 500 Da. The large inhibitory compounds 45375808 and 9806452 were selected and expected to tightly bind to CsSBAT. These four inhibitor candidates revealed least toxicity that may enhance their druggability. Collectively, four compounds were proposed as putative inhibitors of CsSBAT, deserving further *in vitro* and *in vivo* evaluation toward anthelminthic development.

**Declarations**

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**Ethics approval and consent to participate**

Not applicable.

**Consent for publication**

Not applicable.

**Availability of data and material**

The data supporting the conclusions of this article are included in the article.

**Competing interests**

The authors declare that they have no competing interests.
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Authors contributions

SJH and WGY conceived of and designed the experiments. WGY and FD analyzed the data. WGY contributed to utilizing the analysis tools. WGY and JHP wrote the paper. WGY and SJH reviewed the paper. All authors read and approved the final version of the manuscript.

Author details

1 Department of Medical Environmental Biology, Chung-Ang University College of Medicine, Seoul 06974, Korea

2 Department of Parasitology, School of Biology and Basic Medical Sciences, Medical College, Soochow University, Suzhou, Jiangsu, PR China

3 Department of Convergence Medicine, University of Ulsan College of Medicine and Asan Institute for Life Sciences, Asan Medical Center, Seoul 05505, Korea

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**Abbreviations**

ASBT apical sodium-dependent bile acid transporter

BESP bile salt export pump

IF inward-facing

LD<sub>50</sub> median lethal dose

LogP lipophilicity

Mr molecular weight

MRP multidrug resistance-associated protein

NTCP Na+-taurocholate cotransporting polypeptide

OF outward-facing

PATD polyacrylic acid–tetraDOCA conjugate

SBAT sodium-bile acid co-transporter

SBDD structure-based drug discovery

TCH taurocholate

**Figures**
Figure 1

Strategy for structure-based virtual screening
Figure 2

Conformation change of CsSBAT accompanied with transport of Na+ and taurocholate. Both OF-CsSBAT (A) and IF-CsSBAT (B) are colored in tan. OF-YfASBT (A) is colored in pink and IF-NmASBT (B) in blue. ASBTs presenting identical conformation are superposed. Tunnel leading to the TCH-binding site is visualized in yellow eclipse. Residues forming TCH-binding cavity are depicted as green spheres in CsSBAT. Other residues participating in TCH-binding site are provided at each side. Na+ ions are in purple circle (Na1 and Na2). Taurocholate (TCH) is depicted as maroon pentagon.
Figure 3

Na\(^+\) and taurocholate-binding sites of CsSBAT. All ligands and interaction were predicted using COACH [45] except for Na3 site (A). Na\(^+\)-binding sites (B, C), Na1 site (B), and Na2 site (C). Putative Na3 site (D) as suggested by Alhadeff et al. [64]. The consensus residues forming the binding site are presented in stick mode and labeled. Side-chain oxygen and hydrogen atoms are indicated in red and white, respectively. Na\(^+\) ion is depicted as a purple ball and taurocholate substrate as stick mode.
A Venn diagram presents compounds interacting with two SBATs and two ASBTs. Ellipse represents compounds screened against each of the four ASBTs. Gray area indicates compounds, which interact with either OF-CsSBAT or IF-CsSBAT. Black area depicts compounds that interact with both CsSBATs, but not with other ASBTs.
Docked pose of two compounds with IF-CsSBAT and OF-CsSBAT following Lipinski’s rule of five. Binding modes of compounds were obtained from the MTiOpenScreen [47]. In each panel, left box ribbon and stick indicate a compound placed in the active site of OF-CsSBAT (A and C) and IF-CsSBAT (B and D). Right panel generated by LigPlot+ v1.4.3 [54] indicates the schematic interactions of CsSBAT with 124948115 (A and B) and 49734421 (C and D). The compound and residues with hydrophobic interactions are visualized with 2D diagrams. The residues contributing to hydrogen bonds are colored green. Hydrogen bond is indicated as a lengthy broken line. Black arrow indicates compound movement.
Docked poses of large compounds with OF-CsSBAT and IF-CsSBATs. The binding modes of compounds were obtained from the AutoDock Vina v1.1.2 [50]. Left (A, C, E, G, and I) and right (B, D, F, H, and J) panels represent OF-CsSBAT and IF-CsSBAT, respectively. The schematic interactions of CsSBAT with 4537808 (A and B), 9806452 (C and D), 441243 (E and F), 4701 (G and H), and 92727 (I and J) were generated using LigPlot+ v1.4.3 [54]. The compound and residues with hydrophobic interactions are visualized with 2D diagrams. The residues contributing to hydrogen bonds are colored green. Hydrogen bond is indicated as a broken line with length. Amino acid residues involved in the hydrophobic interactions are marked with red spoked arcs. Black arrow indicates compound movement.
Figure 7

Docked poses of large compounds with OF-CsSBAT and IF-CsSBATs. The binding modes of compounds were obtained from the AutoDock Vina v1.1.2 [50]. Left (A, C, E, G, and I) and right (B, D, F, H, and J) panels represent OF-CsSBAT and IF-CsSBAT, respectively. The schematic interactions of CsSBAT with 4537808 (A and B), 9806452 (C and D), 441243 (E and F), 4701 (G and H), and 92727 (I and J) were generated using LigPlot+ v1.4.3 [54]. The compound and residues with hydrophobic interactions are visualized with 2D diagrams. The residues contributing to hydrogen bonds are colored green. Hydrogen bond is indicated as a broken line with length. Amino acid residues involved in the hydrophobic interactions are marked with red spoked arcs. Black arrow indicates compound movement.

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