Short communication

Structure-based virtual screening approach to identify novel classes of PTP1B inhibitors

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ABSTRACT

Discovery of protein tyrosine phosphatase 1B (PTP1B) inhibitors has been actively pursued with the aim to develop therapeutics for the treatment of type 2 diabetes and obesity. We have been able to identify 9 novel PTP1B inhibitors by means of a computer-aided drug design protocol involving virtual screening with docking simulations under consideration of the effects of ligand solvation in the binding free energy function. Because the newly discovered inhibitors are structurally diverse and reveal a significant potency with IC_{50} values lower than 50 μM, all of them can be considered for further development by structure–activity relationship studies. Structural features relevant to the interactions of the newly identified inhibitors with the active-site residues of PTP1B are discussed in detail.

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1. Introduction

Protein tyrosine phosphatase (PTP) was identified 2 decades ago by the purification of PTP1B from human placenta [1]. Subsequent human genome sequencing revealed the presence of more than 100 human PTPs [2]. As can be envisaged by the importance of PTPs in the cellular function, many of the PTPs are implicated in a variety of human diseases including diabetes, obesity, autoimmune diseases, infectious diseases, inflammation, cancer, osteoporosis, and neurodegeneration [3–5]. Among them, PTP1B is known to be associated with type 2 diabetes and obesity [6]. As an enzyme dephosphorylating the insulin receptor (IR) in skeletal muscle and liver, and Jak2 in hypothalamus, PTP1B regulates IR and leptin signaling, eventually acting as an important mediator for control of blood glucose levels and body weight [7–10]. When the PTP1B gene was disrupted or its expression suppressed in mice, reduced blood glucose levels and improved insulin responsiveness were observed in normal and diabetic mice [11–13]. PTP1B inhibition or a reduction of its cellular abundance in mice resulted in increased

sensitivities to leptin and insulin, and has also exhibited a protective effect against diet-induced obesity [11,12,14]. On the basis of these observations, the inhibition of PTP1B has emerged as an attractive therapeutic strategy to treat type 2 diabetes and obesity, which has in the last decade motivated research in this field.

As recently reviewed in a comprehensive fashion [15,16], a large number of PTP1B inhibitors have been developed over the last decade in an effort to design potent and selective compounds as drug candidates. In the design of PTP inhibitors, the primary concern is the construction of a chemical structure that can mimic the phosphotyrosyl functional group to tightly bind in the highly conserved and polarized active site of PTPases. The previously used pTyr mimetics include methylenephosphonate, malonate, sulfate, carboxylate, and difluoromethylenephosphonate groups. Unfortunately, most of these multiple-charged phosphate-mimicking components have proven difficult to develop into effective drugs due to their low cell permeability and bioavailability. Therefore, new potential drug candidates targeting PTP1B need to be developed. Very recently, a selective noncompetitive PTP1B inhibitor, trodusquemine, developed by Genaera, proceeded to phase I clinical trials with promising preclinical and early clinical results as both an appetite suppressant and a hypoglycemic and hypocholestrolemic agent.

In the present study, we identified novel classes of PTP1B inhibitors by means of a structure-based drug design protocol involving virtual screening with docking simulations and an *in vitro* enzyme assay. The characteristic feature that discriminates our

virtual screening approach from the others lies in the implementation of an accurate solvation model in calculating the binding free energy between PTP1B and putative ligands, which would have an effect of increasing the hit rate in the enzyme assay [17,18]. It will be shown that the docking simulation with the improved binding free energy function can be a useful tool for elucidating the activities of the identified inhibitors, as well as for enriching the chemical library with molecules that are likely to have desired biological activities.

2. Materials and methods

2.1. Materials

The docking library for PTP1B comprising about 85,000 compounds was constructed from the latest version of the chemical database distributed by Interbioscreen (http://www.ibscreen.com) containing approximately 30,000 natural and 320,000 synthetic compounds. Chemicals for the enzyme assay were purchased from Sigma (St. Louis, U.S.A.). A substrate, p-nitrophenyl phosphate (pNPP), for the PTP assay was obtained from Sigma in the di(Tris) salt form. The native form of PTP1B was expressed in an *Escherichia coli* expression system, and purified as described previously [19]. Absorbances were measured using Novaspec-II spectrophotometer (Amersham Pharmacia) or DU 650 spectrophotometer (Beckman Coulter).

2.2. Virtual screening of PTP1B inhibitors

The 3-D coordinates in the X-ray crystal structure of PTP1B complexed with an N-phenyloxamate inhibitor (PDB code: 1Q1M) [20] were selected as the receptor model in the virtual screening with docking simulations. After removing the ligand and solvent molecules, hydrogen atoms were added to each protein atom. Special attention was paid to assign the protonation states of the ionizable Asp, Glu, His, and Lys residues in the X-ray structure of PTP1B. The side chains of Asp and Glu residues were assumed to be neutral if one of their carboxylate oxygens pointed towards a hydrogen-bond accepting group including the backbone aminocarbonyl oxygen at a distance within 3.5 Å, a generally accepted distance limit for a hydrogen bond of moderate strength [21]. Similarly, the lysine side chains were assumed to be protonated unless the NZ atom was in proximity of a hydrogen-bond donating group. The same procedure was also applied to determine the protonation states of ND and NE atoms in His residues.

The compounds in the docking library were selected with the drug-like filters that select only the compounds with physicochemical properties of potential drug candidates [22] and without reactive functional group(s). As a consequence of this filtering, the number of compounds decreases from 350,000 in the chemical database to 85,000 in the docking library. All of the compounds included in the docking library were then subjected to the Corina program to generate their 3-D atomic coordinates, followed by assignment of Gasteiger–Marsilli atomic charges [23]. We used the AutoDock program [24] in the virtual screening of PTP1B inhibitors because the outperformance of its scoring function over those of the others had been demonstrated in several target proteins [25]. AMBER force field parameters were assigned for calculating the van der Waals interactions and internal energy of a ligand as implemented in the AutoDock program. Docking simulations with AutoDock were then carried out in the active site of PTP1B to score and rank the compounds in the docking library according to their calculated binding affinities.

In the actual docking simulation of the compounds in the docking library, we used the empirical AutoDock scoring function

improved by the implementation of a new solvation model for a compound. The modified scoring function has the following form:

$$
\begin{aligned}
\Delta G_{\text{bind}}^{\text{aq}} &= W_{\text{vdW}} \sum_{i=1} \sum_{j>i} \left(\frac{A_{ij}}{r_{ij}^{12}} - \frac{B_{ij}}{r_{ij}^{6}} \right) + W_{\text{hbond}} \sum_{i=1} \sum_{j>i} E(t) \\
&\times \left(\frac{C_{ij}}{r_{ij}^{12}} - \frac{D_{ij}}{r_{ij}^{10}} \right) + W_{\text{elec}} \sum_{i=1} \sum_{j>i} \frac{q_i q_j}{\varepsilon(r_{ij}) r_{ij}} + W_{\text{tor}} N_{\text{tor}} \\
&+ W_{\text{sol}} \sum_{i=1} S_i \left(\text{Occ}_i^{\text{max}} - \sum_{j>i} V_j e^{-(r_{ij}^2/2\sigma^2)} \right),
\end{aligned}
\tag{1}
$$

where W_{vdW}, W_{hbond}, W_{elec}, W_{tor}, and W_{sol} are the weighting factors of van der Waals, hydrogen bond, electrostatic interactions, torsional term, and desolvation energy of inhibitors, respectively. r_{ij} represents the interatomic distance, and A_{ij}, B_{ij}, C_{ij}, and D_{ij} are related to the depths of the potential energy well and the equilibrium separations between the 2 atoms. The hydrogen bond term has an additional weighting factor, $E(t)$, representing the angle-dependent directionality. A cubic equation approach was applied to obtain the dielectric constant required in computing the interatomic electrostatic interactions between PTP1B and a ligand molecule [26]. In the entropic term, N_{tor} is the number of sp^3 bonds in the ligand. In desolvation terms, S_i and V_i are the solvation parameter and the fragmental volume of atom i [27], respectively, while $\text{Occ}_i^{\text{max}}$ stands for the maximum atomic occupancy. In the calculation of molecular solvation free energy term in Eq. (1), we used the atomic parameters recently developed by Kang et al. [28] because those of the atoms other than carbon were unavailable in the current version of AutoDock. This modification of the solvation free energy term is expected to increase the accuracy in virtual screening because the underestimation of ligand solvation often leads to the overestimation of the binding affinity of a ligand with many polar atoms [17].

The docking simulation of a compound in the docking library started with the calculation of the three-dimensional grids of interaction energy for all of the possible atom types present in chemical database. These uniquely defined potential grids for the receptor protein were then used in common for docking simulations of all compounds in the docking library. As the center of the common grids in the active site, we used the center of mass coordinates of the docked structure of the probe molecule, N-phenyloxamate, whose binding mode in the active site of PTP1B had been determined with X-ray crystallography and reported in PDB [20]. The calculated grid maps were of dimension $61 \times 61 \times 61$ points with the spacing of 0.375 Å, yielding a receptor model that includes the atoms within 22.9 Å of the grid center. For each compound in the library, 10 docking runs were performed with the initial population of 50 individuals. Maximum number of generations and energy evaluation were set to 27,000 and 2.5×10^5, respectively.

2.3. Enzymatic assay and IC$_{50}$ determination

The compounds identified by virtual screening were tested for their ability to inhibit PTP1B, using pNPP as the substrate. The PTP1B activities were assayed in buffer A (100 mM Hepes, 5 mM EDTA, pH 7.0) at 37 °C in a final pNPP concentration of 2 mM. For the assay, the enzyme was diluted to a proper concentration with enzyme dilution buffer (25 mM Hepes, 5 mM EDTA, 1 mM DTT, and 1 mg/mL bovine serum albumin, pH 7.3). The concentration of PTP1B in the assay mixture was 2.1 μg/mL. A typical 50-μL reaction system contained 5 μL pNPP, 5 μL enzyme, 5 μL inhibitor dissolved in DMSO, 10 μL 5× buffer A, and 25 μL H$_2$O. After the mixture, without pNPP, had been incubated at 37 °C for 10 min, the enzyme