



Exploration of Novel Inhibitors for Class I Histone Deacetylase Isoforms by QSAR Modeling and Molecular Dynamics Simulation Assays

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Abstract

Histone deacetylases (HDAC) are metal-dependent enzymes and considered as important targets for cell functioning. Particularly, higher expression of class I HDACs is common in the onset of multiple malignancies which results in deregulation of many target genes involved in cell growth, differentiation and survival. Although substantial attempts have been made to control the irregular functioning of HDACs by employing various inhibitors with high sensitivity towards transformed cells, limited success has been achieved in epigenetic cancer therapy. Here in this study, we used ligand-based pharmacophore and 2-dimensional quantitative structure activity relationship (QSAR) modeling approaches for targeting class I HDAC isoforms. Pharmacophore models were generated by taking into account the known IC₅₀ values and experimental energy scores with extensive validations. The QSAR model having an external R² value of 0.93 was employed for virtual screening of compound libraries. 10 potential lead compounds (C1-C10) were short-listed having strong binding affinities for HDACs, out of which 2 compounds (C8 and C9) were able to interact with all members of class I HDACs. The potential binding modes of HDAC2 and HDAC8 to C8 were explored through molecular dynamics simulations. Overall, bioactivity and ligand efficiency (binding energy/non-hydrogen atoms) profiles suggested that proposed hits may be more effective inhibitors for cancer therapy.

Introduction

Histone Acetyltransferases (HATs) and HDACs regulate the acetylation and deacetylation events of small alkaline histones associated with DNA double helical structure [1,2]. Interactions of positively charged amino-terminal tails of histones with negatively charged phosphodiester backbones of DNA results in chromatin compaction [3,4]. The associated conformational changes which occur due to acetylation of lysine residues result in chromatin remodeling. Thus, HATs mediated acetylation promotes chromatin relaxation by loosening the packed histones and DNA, thereby facilitating the accession of transcription factors to bind to respective DNA templates [5–7]. In contrast to acetylation, compactness of nucleosome units and controlled gene



expression as a result of deacetylation is mediated by HDACs [1]. Histones play crucial roles in stabilizing the heritable epigenetic changes in gene activity and expression [2]. Any disturbances in these functions may lead to the abnormal expression of genetic material that may cause fatal diseases like diabetes and cancer [8,9].

The critical roles of HATs and HDACs in tumor progression, cardiac and brain disorders have been analyzed in many studies [10-13]. In human, at least 18 HDACs have been identified that are grouped into four classes and comprise of two major categories: Zn⁺² and NAD⁺ dependent [14]. This classification is based on the structural, functional and phylogenetic analysis of HDACs [15]. A detailed overview of cellular compartments and HDAC involvements in diverse biological processes is given in Table 1. Class I comprises of HDAC1-3 and 8 [16-18] which controls many functional and regulatory mechanisms [19]. Importantly, implication of class I HDACs has been monitored in hematological malignancies, resulting in differentiation and proliferation abnormalities of myeloid cells [20]. In tumor progression, most prevailing alteration is linked to abnormal expressions of HDACs [21]. Aberrant expression pattern of HDAC1 has been observed in prostate, gastric, breast and colon [22-25] cancers. Similarly, onset of gastric, cervical and colorectal carcinoma [26-28] is associated with increased expression of HDAC2. Abnormal expression patterns of class I HDAC members are also evidenced in cell proliferation and migration during ovarian and breast carcinomas [29]. In these tumor cells, down-regulation of e-cadherin is associated with HDAC3 overexpression. Unbalanced expressions of HDAC members also result in Acute Promyelocytic Leukemia (APL) such as lymphoblastic APL and non-Hodgkin's lymphomas [28,30].

Table 1. Classification and biological roles of HDACs.

Classes	Cofactors	HDACs	Cellular Locations	Biological Processes	References
Class I	Zn ⁺²	dependent	1, 2, 3, 8	Nucleus, Cytoplasm, Transcriptional repressor complex, Spindle microtubule, Replication fork	Cell cycle
		regulation, Cell	differentiation, DNA damage response, Epidermis development, Regulating cardiac myocyte proliferation on the course of cardiac development	[<u>16–18]</u>	
Class II	Zn ⁺²	dependent	4, 5, 6, 7, 9, 10	Cytoplasm, Nucleus, Neuromuscular junction, Golgi apparatus, Cytosol caveola	Regulation of
		transcription and cell	differentiation, Regulation of cardiac muscle contraction, Inflammatory response, Nervous system development, Heart development, Protein polyubiquitination, Response to toxic and organic substances, Macroautophagy, Vasculogenesis	[<u>1</u> , <u>17</u>]	
Class III	NAD ⁺	dependent	Sirtuins SIRT1-SIRT7	Nucleus, Cytoplasm	Histone
			deacetylation, Regulation of phosphorylation, Regulation of double-strand break repair via homologous recombination, DNA repair mechanism	[16]	
Class IV	Zn ⁺²	dependent	11	Nucleus	
		Transcription, DNA- dependent chromatin modification, Histone	Deacetylation	[16]	



Another mechanism of HDAC-mediated tumor onset is due to transcriptional repression of tumor suppressor genes and their aberrant recruitments to promoter regions. In recent years, many studies potentiate the ways of targeted HDAC inhibition in the context of tumor control [31,32]. Transcriptional activation of tumor suppressor genes by the inhibition of HDAC activity is considered as an ideal and innovative strategy. To date, several HDAC inhibitors (HDACi) like hydroxamic acids, benzamides, short chain carbolic acids, and cyclic tetrapeptides have been characterized *in vitro* and *ex vivo* for various cancers [33–36]. Hydroxamic acids (hydroximates) include vorinostat (SAHA), belinostat (PXD101), panobinostat (LBH589), dacinostat (LAQ824), givinostat (ITF-2357) and trichostatin A (TSA). Benzamides are the derivatives of benzoic acid which include entinostat (MS-275), p-N-acetyldinaline (CI-994), mocetinostat (MGCD0103) and SK-7041. Short chain carboxylic acids (aliphatic acids) are relatively weak inhibitors which comprise of valproic acid or VPA, butyrate and sodium phenyl butyrate. Cyclic tetrapeptides or cyclic peptides are structurally complex molecules which consist of romidepsin (depsipeptide), apicidin (OSI-2040), trapoxin A and trapoxin B (cyclic hydroxamic acid) [37,38]. Currently, these inhibitors are under investigations in clinical trials [39-44]. In cancer patients, functional activity of HDACi is mediated extrinsically as well as intrinsically due to cell cycle arrest at G1/ G2 phases and up-regulation of apoptosis [31,45]. Sensitivity of actinotheraphy and chemotherapy for cancers is also enhanced by HDAC inhibition [46].

In silico drug designing strategies are extensively applied for the identification of novel inhibitors through modeled HDAC structures and their pharmacophore-based studies [47–53]. However, these inhibitors are unique in terms of HDAC targeting actions. Here, through comparative ligand-based approaches, several novel candidate hits were proposed and added in the list of HDACi which are able to target entire class I of HDACs.

Methodology

Dataset

High resolution co-crystallized structures of class I HDACs were retrieved through PDB database [54] with following PDB IDs: 4BKX (HDAC1), 4LXZ (HDAC2), 4A69 (HDAC3) and 1T64 (HDAC8) (Table A in S1 File). These structures were further refined by adding charges and missing residues through *dock prep* module of UCSF Chimera 1.10 [55]. Binding and catalytic site details of class I HDACs were extracted through PDBSum [56]. Structural superimposition was performed by UCSF Chimera 1.10 [55]. Additionally, comparative binding mode of class I HDACs has been reviewed through literature [47,48,53].

Pharmacophore/QSAR Model Generation

In order to generate the ligand-based quantitative structure-activity relationship (QSAR) and pharmacophore models, 16 known inhibitors (Table B in S1 File) were selected. These known HDAC inhibitors were collected through literature survey [57–73] and classified on the basis of their inhibition potential values for class I HDACs. Chemical structures and bioactivities of these inhibitors were extracted through EMBL-EBI ChEMBL database [74], while their 2D/3D molecular descriptors values were retrieved through ChemAxon's Chemicalize [75] and Molinspiration [76]. Molecular dockings of these inhibitors were performed against class I HDACs by AutoDock VINA with default parameters i.e. exhaustiveness = 8 and energy range = 3 [77].

The selection of training set compounds was based on bioactivity and binding affinity attributes of docked protein-inhibitor complexes. The best docked conformations of inhibitors having least IC_{50} values and high binding abilities were filtered out for pharmacophore generation. In order to design 2D QSAR model, IC_{50} values of training set compounds against HDAC1, 2,



3 and 8 were converted into pIC_{50} values and average pIC_{50} values were calculated for each of the training set compound.

Test data set was created to evaluate the quality of pharmacophore hypotheses. Known selective and non-selective synthetic molecules were gathered through ChEMBL [74] which were utilized in human assays [68,69,71,73,78–109] to generate a dataset of 71 compounds (Table C in S1 File) and their experimental IC50 values were assessed. Compounds having weak inhibition potential (IC50 > 2500 nM) for class I HDACs were grouped in decoy set, whereas high inhibitory compounds (IC50 < 2500 nM) were placed in the active dataset. By implying the above criteria, 41 compounds were found to be active and 30 were declared as decoys. 3D generation and optimization of data set compounds was performed by ACD ChemSketch [110].

To generate pharmacophore models, training set hits were employed in the ligand-based module of the LigandScout 3.0 [111]. Pharmacophoric sites such as hydrogen bond donor (HBD), hydrogen bond acceptor (HBA), hydrophobic sites, aromatic ring and positive and negative ionizable groups were carefully characterized. To incorporate the associated features of selected compounds, merge feature model generation and atom overlap scoring function of LigandScout 3.0 was applied. Subsequently, descriptor selection analysis was performed through forward selection (FS), backward elimination (BE) and stepwise selection (SS), the classical methods of variable selection [112]. All unique descriptors signifying the model were selected, while excluding the non-significant descriptors. QSAR model was created by employing multiple linear regression (MLR) technique [113]. Regression module of the IBM statistical package for social sciences (SPSS) version 22 [114] was used to create the regression model by employing Eq 1.

$$y_i = \beta_0 + \beta_1 x_i + \varepsilon_i, \quad i = 1, \dots, n$$
 (1)

Where, y (pIC₅₀) is the dependent variable, x is the independent variable, β is the coefficient and ε is the error value. QSAR and pharmacophore models were statistically evaluated to check their reliabilities prior to implementation.

Top five models exhibiting significant statistical values were scrutinized as 3D query for virtual screening. Validated 3D pharmacophore models were screened against a total of 48,386 compounds (2,601 of Aurora [$\underline{115}$] and 45,785 of Princeton [$\underline{116}$] libraries) to prey novel drug targets. The novel hits were selected for docking analysis on the basis of high query fit values relative to pharmacophore models.

Molecular docking and dynamic simulation assays

Molecular docking analyses of highly ranked compounds were performed for class I HDACs through above mentioned parameters of AutoDock VINA [77]. Subsequently, energy scores and HDAC bound poses of these compounds were carefully evaluated. To study the structural dynamics of receptor-ligand complex, molecular dynamics (MD) simulations of HDAC2 and HDAC8 were performed. Gromacs 4.5.5 [117] and GROMOS96 43a1 force field with SPC water model was used for MD simulations. Subsequently, energy minimization was performed by steepest descent method and appropriate counter ions were added to neutralize the system [118,119]. Finally, a 12 ns MD simulation run was performed at constant temperature (300 K) and pressure (1 atm). GROMACS analysis tools were used for the analysis of MD simulation trajectories (118). The stability of secondary structure elements, conformational changes and interactions were assessed by computing root mean square deviation (RMSD), root mean square fluctuation (RMSF) and hydrogen bonds obtained throughout MD trajectories. For RMSD, RMSF, gyration and hydrogen bonds g_rms, g_rmsf, g_gyrate and g_hbond modules



of Gromacs were applied, respectively. RMS cluster distribution of HDAC2 and HDAC8 backbone was computed using g_cluster module with nearest neighbor method. Energy values along the MD trajectories were calculated by Eq 2.

$$E = E_{bond \ stretch} + E_{angle \ bend} + E_{rotation} + E_{vander \ waals} + E_{electrostatic}$$
 (2)

All simulations were carried out through an open SUSE 11.2 system with Intel(R) core (TM) i5-2300 CPU containing Linux 2.6.31.5–0.1 operating system. The binding patterns were carefully monitored through Accelrys Discovery Studio 4 [120].

ADMET (Absorption, Distribution, Metabolism, Excretion and Toxicity) properties were predicted through AdmetSAR [121] server to hypothetically depict the positive and negative biological effects of compounds. Bioavailability, Rule of Five (ROF), lead-likeness and other filters were assessed using Chemicalize [75]. Toxicity parameters such as mutagenic and tumorigenic effects of selected compounds were evaluated through OSIRIS [122].

Expected IC₅₀ values of compounds were predicted through QSAR model. Physiochemical descriptors of compounds were extracted by Chemicalize [75] and Molinspiration [76] servers. To avoid any redundancy issue, predicted compounds were compared against small molecule library available in ChEMBL [74] database, based on \geq 90% similarity threshold. Synthetic accessibility analysis of predicted hits was performed through SYLVIA [123] tool.

Results

Binding site analysis

Binding sites of class I HDAC family members were evaluated by structural comparisons (Fig 1). Co-crystallized structures of HDAC1, 2, 3 and 8 showed structural similarities with an RMSD value of 0.5\AA (Fig 1A). Zn⁺² ions were coordinated by one HIS and two ASP residues of individual HDACs (Fig 1B). These catalytic residues resided within the 4Å region of surface forming the tunnel shaped pocket.

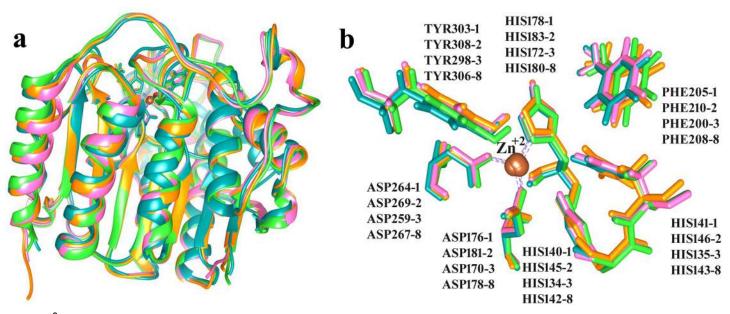


Fig 1. Zn⁺² dependent class I HDACs. (a) Structural superimposition of HDACs. (b) Binding sites of class 1 HDACs. HDAC1, 2, 3 and 8 are shown in orange, pink, green and blue, respectively. Zn⁺² is shown in brown color.



Pharmacophore-based 2D QSAR modeling

In order to select training set compounds, binding affinities, IC_{50} and molecular descriptors values of known class I HDAC inhibitors were monitored (Tables B, D and E in S1 File). These known hits included MS-275, LBH-589, LAQ-824, Trichostatin A, Saha, Belinostat, Oxamflatin, Pyroxamide, Mocetinostat, and Scriptaid with IC_{50} values ranging from 5–300 nM. 5–7 pharmacophore models were generated based on the information of docking poses of training set compounds and their pharmacophore fit values were evaluated. The ideal pharmacophore models with high query fit values were shown in Figure A in S1 File. These models exhibited good pharmacophoric features such as hydrogen bond donors, hydrogen bond acceptors, hydrophobic sites and negative ionizable groups (Zn^{+2} binding locations). QSAR regression model was generated using 5 descriptors of training set compounds, as shown in Eq 3. Their pIC₅₀ values were shown in Table F in S1 File.

$$pIC_{50} = 14.48(\pm 3.8) - 0.13(\pm 0.06) * Mol.polarizability + 0.06(\pm 0.04) * logP + 0.17(\pm 0.04) * Topological Polar Surface Area (TPSA) - 2.54(\pm 0.5) * Hy.Bond Acceptor - 2.57(\pm 1.21) * Balaban Index(3)$$

To determine the reliability and predictability of models, several statistical parameters (Table G in S1 File) were applied which showed positive results. Higher sensitivity and specificity values, percent of active yield (%Y), enrichment factor and Guner-Henry score depict the performance quality of pharmacophore model [124] (Fig 2A). Our generated models showed high sensitivity and specificity values; models 1 and 2 resulted in 80% and 75% retrieval of active compounds, respectively. Scores of other statistical parameters were comprised of 80% score of number of active percent of yields (%Y) and G.H scores of 0.7 and 0.65 for models 1 and 2, respectively. Statistical evaluation of QSAR model included goodness of fit i.e. R^2 (correlation coefficient) and adjusted R^2 (goodness of fit) values (coefficient of determination). The R^2 value (0.93) and adjusted R^2 (0.84) were close to 1 and standard error value was close to 0 (Table 2). External validation was performed by predicting pIC₅₀ values of training set compounds through generated model and cross-validated against the given activity values (Fig 2B). Observed IC_{50} values of training set compounds were comparable with the predicted values

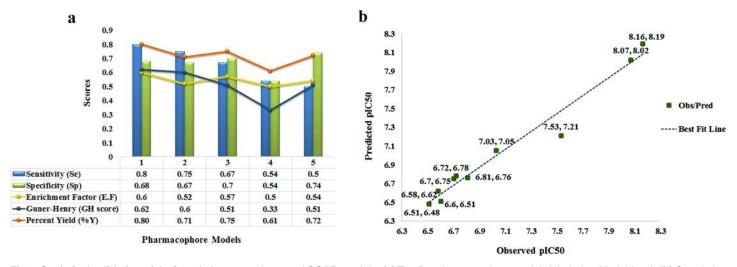


Fig 2. Statistical validation of designed pharmacophore and QSAR models. (a) Top five pharmacophore models labeled as Model (1–5). (b) Correlation analysis of QSAR regression model. Green dots depict the observed vs predicted IC₅₀ values and correlation best fit dotted line is shown in blue.



Table 2. Statistical parameters for regression model.

N	R	R ²	Adjusted R ²	Std. Error	Significance
10	0.96	0.93	0.84	0.24	0.02

and residual error values were \leq 0.5. ANalysis of VAriance (ANoVA) test [125] also verified the model with a p-value of 0.02.

Virtual screening

Pharmacophore model was incorporated to the virtual screening of compound libraries and about 2000 hits were screened out with high to low query fit values. 81 top scoring hits were selected for docking and drug-like analysis. 16 novel hits (1–16 in Table H in S1 File) exhibited interactions with specified binding sites of class I HDACs. These novel compounds were further exploited for the analysis of steric and physiochemical properties (Table I in S1 File) and 10 compounds (C1-C10) were identified as potential inhibitors for class I HDACs. The 2D structures and binding profile of these compounds are listed in Fig 3 and Table J in S1 File indicates their IUPAC names and SMILE codes. These compounds (C1-C10) exhibited interactions with Zn⁺² and binding pocket residues of class I HDACs. C8 and C9 showed class specific binding (Fig 4), while other hits were specific for individual HDACs (Figures B-E in S1 File). Binding pattern of new drug-like compounds coincides with the experimental known binding of existing inhibitors. Bound complexes of SAHA with HDAC2 and Trichostatin A with HDAC8 were shown in Figure F in S1 File which depicted that these inhibitors have similar interaction mode as described for the newly identified inhibitors (Figures B-E in S1 File). Interacting residues and distances of proposed compounds (C1-C10) were also in good agreement to the experimental results.

ADMET analysis, binding contributions and physiochemical properties suggested that predicted hits (C1-C10) (<u>Table 3</u>) may act as more potent inhibitors against class I HDACs. Synthetic accessibility scores (<5) of these hits (<u>Fig 5</u>) further validated these hits. The expected pIC₅₀ values of these hits were predicted by QSAR model (<u>Eq 1</u>) and compared with the known pIC₅₀ values of training set compounds. As evident, hit C7 (pIC₅₀ = 12.5) is the most bioactive compound (<u>Fig 5</u>).

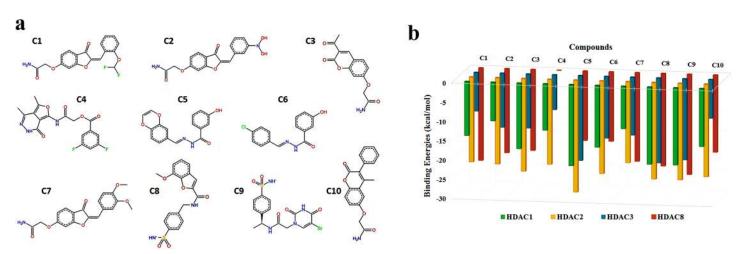


Fig 3. 2D structures and binding energies of compounds C1-C10. (a) 2D structures (b) Binding energy overview.



Recently, the properties like ligand efficiency (LE) or binding energy of ligand per atom [126-128] and lipophilic efficiency (LipE or LLE) [129, 130] have been considered essential for lead optimization [131,132], which rely on both potency and lipophilicity profiles. LE is the ratio of free energy of binding to the number of heavy atoms and was calculated by assuming the standard conditions of aqueous solution i.e. 300K, neutral pH and remaining concentrations of 1M (Eq.4) [133]. To access the lipophilicity of predicted hits, logP (activity/size) values were calculated (Table.4) using Bio-Loom version 1.5 [134] and compared with pIC₅₀ values. LE (Eq.4) [135,133] and LipE (Eq.5) [129] profiles of inhibitors were used to identify the hits with higher activities (Table.4).

$$LE = (1.37/HA) * pIC_{50}$$
 (4)

$$LLE = pIC50 - clogP (5)$$

Size-independent ligand efficiency values (LEScale) of these hits were calculated through fitting the top LE values with heavy atom count through a simple exponential function (Eq.6), as described [127,128]. "Fit Quality" or "FQ" scoring function (Eq.7) is the ratio of LE and

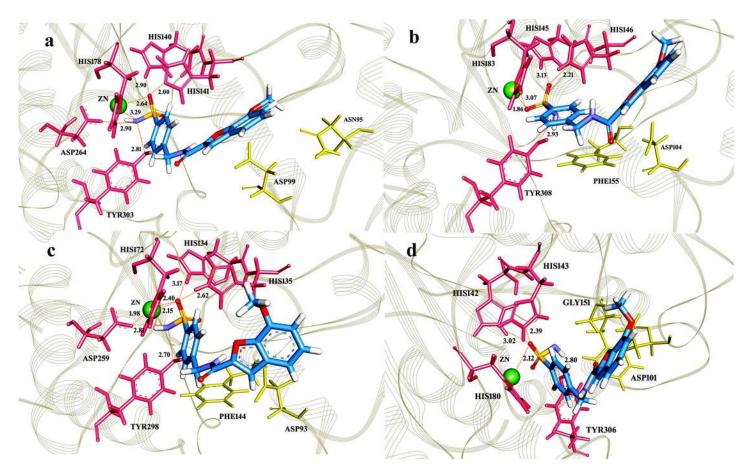


Fig 4. Binding pattern of compound C8 with class 1 HDACs. C8 (7-methoxy-N-((4-sulfamoylphenyl)methyl)-1-benzofuran-2-carboxamide) was shown in blue, whereas hydrogen bonding and hydrophobic residues were shown in pink and yellow, respectively. (a) Hit C8 forms hydrogen bonds with HIS140, HIS148, HIS178, ASP264 and TYR303 of HDAC1; (b) HIS145, HIS146, HIS183 and TYR308 of HDAC2; (c) HIS134, HIS135, HIS172, ASP259 and (d) TYR298 of HDAC3 and HIS142, HIS143, HIS180 and TYR306 of HDAC8. C8 bonding with Zn+2 is shown in green color with an average distance of 2Å. Hydrophobic residues involved in interaction are (a) ASP99, GLY149, PHE150 and GLY301 in HDAC1; (b) PHE155, PHE210, ASP269 and GLY306 in HDAC2; (c) ASP93, PHE144, ASP170, PHE209 and GLY296 in HDAC3 and (d) PHE152, PHE208, MET274 and GLY304 in HDAC8.



Table 3. ADME and Toxicity Analysis.

Compounds	Human Intestinal	CYP450 Inhibition	Blood Brain	Bioavailability	No	No	No
	Absorption	Inhibition	Barrier		Mutagenic	Tumorigenic	Irritant
C1	1	0.72	0.964	+	+	+	+
C2	0.8	0.6	0.7	+	+	+	+
C3	0.9	0.8	0.7	+	+	+	+
C4	0.9	0.6	0.7	+	+	+	+
C5	0.7	0.6	0.7	+	+	+	+
C6	0.9	0.8	0.9	+	+	+	+
C7	1	0.8	0.7	+	+	+	+
C8	1	0.8	0.8	+	+	+	+
C9	0.9	0.8	0.6	+	+	+	+
C10	1	0.7	0.6	+	+	+	+

LEScale which was computed to detect the optimal ligand binding properties of predicted hits.

$$LEScale = 0.104 + 0.65e^{-0.037*HA}$$
 (6)

$$FQ = LE/LEScale (7)$$

Conventionally, clogP <3, LipE > 5 and FQ close to 1 are considered as optimal properties for highly bioactive compounds [127-129]. Clearly, clogP values of C1, C2, C3, C4, C7, C8 and C10 hits ranged in 0.2 to 2.5 ($\underline{\text{Fig 6A}}$), while about 50% compounds exhibited LipE > 5 ($\underline{\text{Table 4}}$). Similarly, FQ scores of these compounds were also in acceptable range ($\underline{\text{Fig 6B}}$), indicating that proposed hits may have better *in vivo* performance based on their ligand binding, potency and lipophilicity profiles.

Molecular dynamics simulation analysis

To elucidate the dynamic behavior of class I HDACs upon binding to inhibitor and to gauge the pattern of system stability, HDAC2 and HDAC8 were subjected to molecular dynamics (MD) simulations. The stability of secondary structure elements and conformational changes were assessed by computing root mean square deviation (RMSD) and root mean square fluctuation (RMSF) plots of values obtained throughout MD trajectories. RMSD is a measure of

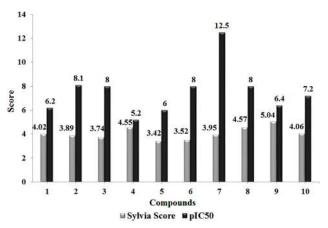


Fig 5. Synthetic accessibility score and expected pIC₅₀ values.



Compounds	pIC ₅₀	HA	clogP	LE	LEScale	LipE	FQ
C1	6.2	25	2.94	0.34	0.36	3.26	0.95
C2	8.1	25	0.81	0.44	0.36	7.29	1.22
C3	8	19	0.27	0.58	0.43	7.73	1.35
C4	5.2	27	1.49	0.26	0.34	3.71	0.76
C5	6	22	4.11	0.37	0.29	1.89	1.28
C6	8	19	3.38	0.58	0.43	4.62	1.35
C7	12.5	26	2.23	0.66	0.35	10.27	1.89
C8	8	25	1.7	0.44	0.36	6.3	1.22
C9	6.4	25	-0.2	0.35	0.36	6.6	0.97
C10	7.2	23	2.5	0.43	0.38	4.7	1.13

stable interaction pattern of docked complex. In case of HDAC2, RMSD trend remained stable for both bound and unbound complexes within the range of 1.5-2Å (Fig.7A). However, In case of apo HDAC8, an increasing trend of RMSD profile (1.5Å to 2.8Å) was observed between 2–4 ns, while in its bound form with C8, system attained lower RMSD values in a range of 1.5Å-2.2Å (Fig.7B). Our analysis indicated that backbone RMSD profiles of HDAC2 and HDAC8 systems remained stable during 12 ns MD runs.

RMSF plots provided insight into the residual fluctuations upon binding to inhibitor (Fig 8). In HDAC2 bound to C8, higher RMSF values were observed for ILE40-THR43 (2–2.7Å), PHE210 (0.8-2Å) and LEU333 (1.5-2Å) residues (Fig 8A), while residues involved in inhibitor binding namely, HIS145, HIS146, ASP181, HIS183, ASP269 and TYR308 were quite stable. In HDAC8-C8 system, fluctuations were more common in ASP87, ASP88, SER193-SER199 and GLY320 residues. However, RMSF profiles of binding site and Zn⁺² coordinated residues (HIS142, HIS143, ASP178, HIS180, ASP267 and TYR306) exhibited a lower trend (Fig 8B). The observed conformational changes occurring in the proximal residues of HDAC2 and HDAC8 binding sites induced more flexibility to accommodate the inhibitor.

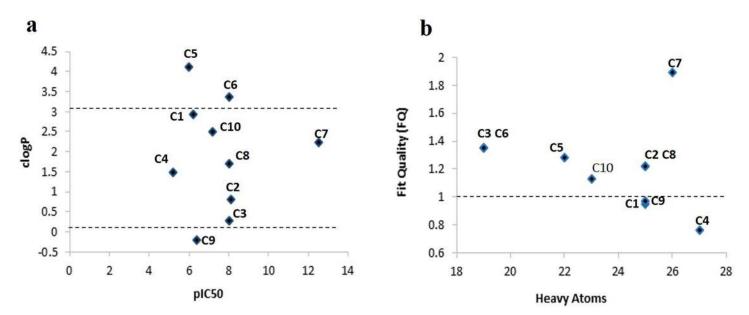


Fig 6. Ligand efficiency analysis of 10 selected compounds (C1-C10). (a) ClogP and (b) Fit Quality graph.



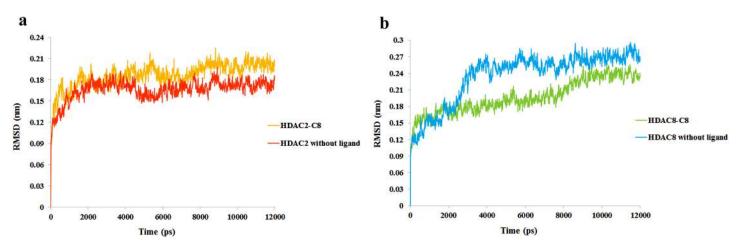


Fig 7. RMSD plot for 12 ns MD simulation. (a) RMSD plot for HDAC2-C8 complex (yellow) and HDAC2 without ligand (red). (b) RMSD plot for HDAC8-C8 complex (green) and HDAC8 without ligand (blue).

The potential energy of system is a measure of its stability. By plotting potential energy as a function of time, we observed that systems were well equilibrated and remained stable throughout MD simulations. The HDAC2-C8 (-74100 kcal/mol) system has shown lower potential energy values compared to HDAC8-C8 (-727500 kcal/mol) complex (Fig 9). The binding characteristics of HDAC2 and HDAC8 with C8 were analyzed through plotting time-dependent intermolecular hydrogen bonds. Compared to HDAC8-C8 system, the intermolecular hydrogen bonds were increased in HDAC2-C8 complex after 2 ns of simulation time, indicating higher interactions (Fig 10). Overall, hydrogen bond interactions remained stable throughout the simulation time. These results substantiated that C8 exhibited more stable binding to HDAC2 compared to HDAC8, which is in good agreement to the distribution of RMSD distances of trajectory conformation pairs. The lower mean RMS values in case of HDAC2-C8 complex displayed the stable nature of system compared to apo HDAC2, while in C8 bound and unbound HDAC8, the pattern of RMS distribution was quite similar (Fig 11). Analysis of radius of gyration (Rg) together with RMSD profile revealed a convergence of Rg values between 1.88 nm to 1.91 nm for HDAC2-C8 complex (Fig 12A). In case of apo HDAC8, Rg

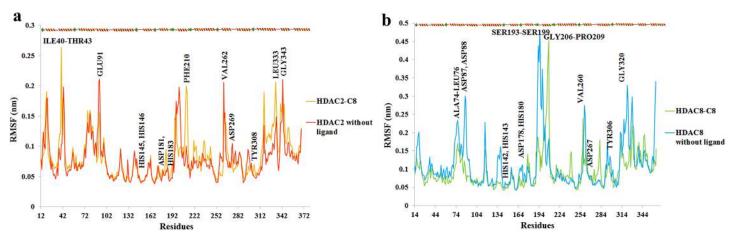


Fig 8. RMSF plot for 12 ns MD simulation. (a) RMSF plot for HDAC2-C8 complex (yellow) and HDAC2 without ligand (red). (b) RMSF plot for HDAC8-C8 complex (green) and HDAC8 without ligand (blue).

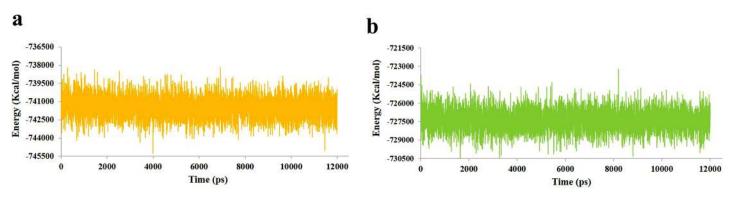


Fig 9. Energy plot for 12 ns MD simulation. (a) Energy plot for HDAC2-C8 complex (yellow). (b) Energy plot for HDAC8-C8 complex (green).

trend was quite different compared to its bound form (Fig 12B). These data indicated that structural transitions in HDAC2 resulted in less tight packing, while in HDAC8, binding of C8 induced more compaction in the structure.

Discussion

Ligand-based pharmacophore modeling, based on the energetic binding values of inhibitors (epharmacophore) is considered as effective tool in rational drug design [53]. Here we retrieved a diverse set of inhibitors for class I HDACs by generating pharmacophore/QSAR models based on QSAR properties of MS-275, LBH-589, LAQ-824, Trichostatin A, Saha, Belinostat, Oxamflatin, Pyroxamide, Mocetinostat, and Scriptaid. The quality of our selected model (R² value of 0.93) was quite high, compared to previously proposed pharmacophore/QSAR models for screening HDAC inhibitors [49–51]. Subsequently, 10 novel hit compounds (C1-C10) were determined through their query fit values and pharmacophoric features they possess. Our current protocol of pharmacophore/QSAR-based virtual screening was quite efficient in predicting the reliable inhibitors as it was based on a subset of selective descriptors which were based on known scaffolds.

Through docking analysis, the candidate hits with binding affinities for all members of class I HDACs were selected for detailed analysis. The structural features of predicted hits were much similar by having a capping region, a linker and a Zn⁺² ion binding region. The predicted hits were derivatives of benzamides, acetamides, carboxamides and hydrazides. Out of these hits, C8 (7-methoxy-N-[(4-sulfamoylphenyl)methyl]-1-benzofuran-2-carboxamide) and C9 2-

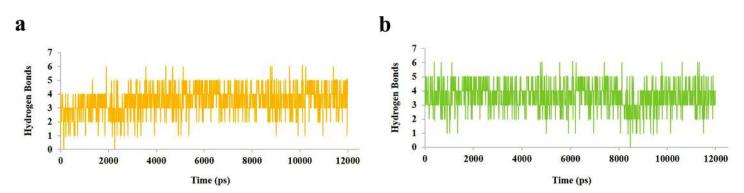


Fig 10. Hydrogen bonds plot for 12 ns MD simulation. (a) Hydrogen bonds for HDAC2-C8 complex (yellow). (b) Hydrogen bonds for HDAC8-C8 complex (green).



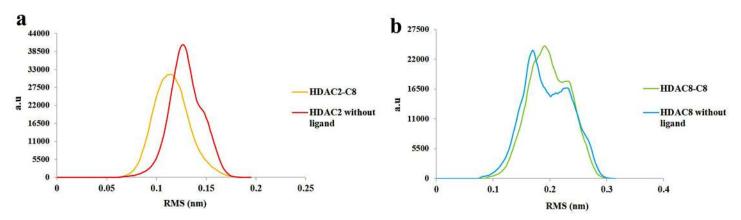


Fig 11. RMS distribution for 12 ns MD simulation. (a) RMS distribution pattern for HDAC2-C8 complex (yellow) and HDAC2 without ligand (red). (b) RMS distribution for HDAC8-C8 complex (green) and HDAC8 without ligand (blue).

(5-bromo-2,4-dioxo-1,2,3,4-tetrahydropyrimidin-1-yl)-N-[1-(4-sulfamoylphenyl)ethyl]acetamide showed ideal binding energetics for HDAC1, 2, 3 and 8 (Fig 3). In C8, carboxamide moiety was involved in hydrogen bonding with Zn⁺² metal ion as well as other active site residues. Due to possessing Zn⁺²-binding amide moieties and stereo selectivity, carboxamides are considered as more potent inhibitors for HDACs [136,137]. Recent experimental evidences support benzofuran-2-carboxamide derivatives for their antitumor and anti-proliferative properties [138]. Similarly, various benzofuran-2-carboxylic acids bearing (chlorometyl) indoline or benzoyl nitrogen as DNA-binding group serve as structural subunits of synthetic analogues of natural antitumor agents such as dystamycin, CC-1065, duocarmycin, and netropsin [139-141]. They also act as adenosine A2A receptor antagonists [142]. Based on the essential value of carboxamides, hit C8 was analyzed in detail for its binding to class I HDAC isoforms. In the catalytic pockets of HDACs, sulfamoylphenyl ring of C8 actively participated in the hydrogen bonding with HIS140, HIS148, HIS178, ASP264 and TYR303 residues of HDAC1; HIS145, HIS146, HIS183 and TYR308 of HDAC2; HIS134, HIS135, HIS172, ASP259 and TYR298 of HDAC3 and HIS142, HIS143, HIS180 and TYR306 of HDAC8 (Fig 4). These interactions were mainly mediated by the conformational readjustments of linker region, while benzofuran ring was uniquely involved in hydrophobic interactions. Consequently, the O-atoms of SO₂ coordinated with metal ion in HDAC hydrophobic pockets. Indeed, study of hydrogen

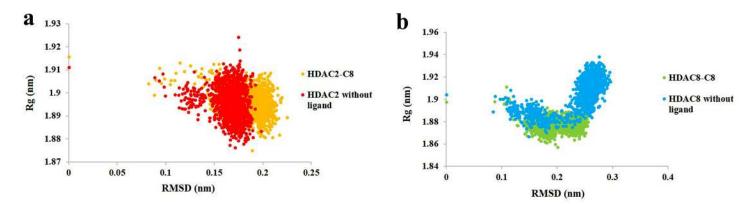


Fig 12. Radius of gyration (Rg) analysis for 12 ns MD simulation. (a) Rg/RMSD plot for HDAC2-C8 complex (yellow) and HDAC2 without ligand (red). (b) Rg/RMSD for HDAC8-C8 complex (green) and HDAC8 without ligand (blue).



bond network between catalytic metal ion and sulfonamide group may be useful to synthesize more potent inhibitors for HDACs.

The proposed aromatic/heterocyclic compounds possessed isoform-specific interactions. For example, crystal structure of HDAC1 exhibited binding with C3, C5 and C6 hits, HDAC2 showed binding with C1-C10, HDAC3 with C5, C6 and C7, while HDAC8 was complexed to C1, C2, C3, C7 and C10. Interestingly, C1, C4, C6 and C9 compounds exhibited halogens as capping groups which indicated that these inhibitors may prefer class I HDACs for binding. The importance of halogenated capping in the selectivity of inhibitors has been well documented. In a series of aroyl pyrrolyl hydroxamide (APHA) compounds screened against maize HD1-B and HD1-A (homologues of mammalian class I and class II HDACs), non-halogenated derivatives showed no selectivity for class I or II [143]. Modification of the capping group region led to 176-fold selectivity for class I HDAC over II. These data underscore that halogens impart class selectivity. However, further analysis is warranted to understand the significance of capping region in parallel to variations in the choice of linker region and metal binding groups in HDACi selectivity.

Moreover, compounds with conserved scaffolds exhibited the comparable binding pattern and energetics. Here, C1, C2 and C7 analogues (group I) possessed similar linker and Zn⁺² ion binding regions with distinguished capping region substituents i.e. difluoromethoxy, carbamoylmethoxy and dimethoxyphenyl, respectively (Fig 3A). Likewise, compounds C5 and C6 (group II) were distinct only in the capping site where the benzodioxin in C5 was replaced by chlorophenyl in C6. (Fig 3A). Energy profiles of group I analogues (C1, C2 and C7) against HDAC isoforms depicted more selectivity for HDAC2 and HDAC8 (Fig 3B). These acetamides derivatives showed similar binding patterns at Zn⁺² containing pockets of HDAC2 and 8 (Figures G and H in S1 File), while at the outward surface of binding tunnel, only C1 and C2 showed binding to the PHE (PHE210 of HDAC2 and PHE208 of HDAC8) and TYR (TYR308 of HDAC2 and TYR306 of HDAC8) residues (Figures G and H in S1 File). Similarly, group II (C5 and C6) hydrazides showed binding selectivity against HDAC1, 2 and 3 (Fig 3B). Metal binding regions of C5 and C6 hits formed hydrogen bonds with active site residues of HDAC enzymes, whereas unlike C6, C5 exhibited binding of benzodioxin with HIS178, HIS183, HIS172 of HDAC1, 2 and 3, respectively (Figures I—K in S1 File). Furthermore, predicted bioactivity analysis of these compounds also showed variability where C7 (group I) having dimethoxyphenyl group and C6 (group II) with chlorophenyl substituent exhibited the highest values (12.5 and 8), respectively. (Fig 5).

MD simulations of C8 with HDAC2 and HDAC8 impressively illustrated the conformational readjustments in the corresponding catalytic sites. Generally, certain conformational changes take place in the vicinity of HDAC catalytic sites to accommodate the inhibitor [144,145]. Binding of C8 was mediated by coordination with metal ion and hydrogen bond acceptors such as HIS145, HIS146, HIS183 and TYR308 in HDAC2 and HIS142, HIS143, HIS180 and TYR306 in HDAC8. These residues showed no structural dynamics during the inhibitor binding. Notably, hydrophobic residues namely ILE40-THR43, PHE210 and LEU333 exhibited more fluctuations in HDAC2 bound to C8 hit. Particularly, PHE210 exhibited the highest peak (0.8-2Å) in RMSF plot. This residue is conserved across the class I HDACs. Previously, an active contribution of PHE210 was detected in the binding of NSC746457 [146] and YF479 inhibitor [147] to HDAC2. Moreover, its involvement is also visible in TSA (Trichostatin A)-HDAC2 crystal structure [148]. The consistent trend in residual conformations suggested an active contribution of PHE210 in the induction of inhibitor binding to HDAC2. In HDAC8, more pronounced fluctuations were observed in ASP87, ASP88, SER193-SER199 and GLY320 residues. However, the residues holding the Zn⁺² ion in the catalytic sites of HDAC2

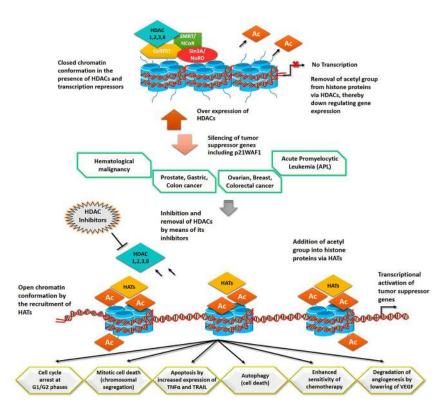


Fig 13. Disease association and inhibition mechanism of class I HDACs. Over expression of HDACs has been perceived in numerous cancers. Binding of drug-like molecules to the catalytic cavity of HDACs eradicates these enzymes from transcription initiation site and stimulates the inclusion of HATs. Acetylation of histones by means of HATs reactivates the transcription of tumor suppressor genes. Expression of these genes controls the abnormal cell growth by cell cycle arrest and mitotic cell death or escalation of apoptosis and autophagy. Tumor suppressor genes enhance the sensitivity of chemotherapy and restrict the process of angiogenesis by decreasing VEGF.

(ASP181, HIS183 and ASP269) and HDAC8 (HIS142, HIS143, ASP178, HIS180, ASP267 and TYR306) were quite stable.

With these structural insights, our nominated drug-like compounds may prove to be operational in combinatorial cancer therapy (Fig 13). Further studies are needed to delineate the inhibitory effects of proposed inhibitors through *in-vitro* and *in-vivo* assays.

Conclusions

We employed QSAR approach to generate models of 10 chemical compounds (C1-C10) which were derivatives of benzamides, acetamides, carboxamides and hydrazides. Pharmacophore models demonstrated the improved values which stated the higher predictability and reliability of screened hits. The predicted hits were tested for their inhibitory effects through docking and MD simulation assays. C8 and C9 targeted all four members of class I HDACs, whereas other compounds showed member specific interactions. The associated conformational changes in the close proximity of binding regions assisted in the interaction of C8 to HDAC2 and HDAC8. RMSD trends for HDAC2 were much similar in both bound and unbound forms, while HDAC8 bound to C8 attained more stability than apo HDAC8. On the basis of *in-silico* binding analysis, these hits may prove as more potent drugs against class I HDACs than previously known analogs to reverse the disequilibrium of acetylation and deacetylation events.



Supporting Information

S1 File. Supporting information. (PDF)

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Author Contributions

Conceived and designed the experiments: SR. Performed the experiments: ZN. Analyzed the data: ZN NA. Contributed reagents/materials/analysis tools: SR. Wrote the paper: SR ZN.

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