

Evaluation of *Cardiospermum halicacabum* leaf compounds against human DihydroOrotate Dehydrogenase: a target for Rheumatoid Arthritis using Structure based Drug Designing

Priya Swaminathan, Lilly Saleena*

Department of Biotechnology, SRM University, Chennai, India.

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ABSTRACT

Rheumatoid arthritis [RA] is a disorder that makes the abnormal immune cells to assail the joints in the body leading to pannus formation and destruction of the synovium. Human DiHydroOrotate DeHydrogenase [hDHODH] inhibition is effective in controlling the proliferation of the abnormal immune cells. Though technology has benefitted science and medicine in many ways, certain indigenous cures are still considered a boom for various diseases. *C. halicacabum* an Indian medicinal plant is widely used as a home remedy for arthritis pain. The novelty of this study relies on the extensive validations performed to obtain robustness of the results. An E-pharmacophore model based on A771726- an experimentally resolved inhibitor of hDHODH was screened against the phytochemicals in the leaf of *C. halicacabum*. These compounds were docked and further validated statistical and ranking methods. Further, flexible docking was performed to understand the optimal pose of the docked structures of Apigenin 7O-glucuronide and Luteolin 7O-glucuronide. These optimal poses were then simulated for 10 ns in a SPC environment which gave a very low RMSD value of 1.5 Å. These results were comparable with the known inhibitor of hDHODH. Thus this study helps to understand and evaluate the different probable inhibitors of hDHODH from *C. halicacabum* and suggests a mode of action for phytochemicals of *C. halicacabum* against RA.

INTRODUCTION

Human DiHydroOrotate DeHydrogenase [hDHODH] is a well studied target for Rheumatoid arthritis [RA] and various other proliferative diseases like multiple sclerosis, cancer and parasitic infections (Munier-Lehmann *et al.*, 2013). It plays a role in *denovo* pyrimidine synthesis for faster proliferation of T cells. The proliferation of T cells is limited when hDHODH is inhibited. Thus the strategy of controlling proliferation by inhibiting hDHODH leads to abatement of symptoms hence leading to decline in the destruction caused by hyperactive immune cells (Fox *et al.*, 1999). Leflunomide is a prodrug known

to inhibit hDHODH. The drug is effective with an IC50 value of 600nm square but prolonged usage generally causes hepatic disorders that are classified as DILI (Drug Induced Liver Injury) (DeLeve and Kaplowitz, 2013). Thus the search for various alternatives that can be effective inhibitors as well as non toxic has been carried out in a number of studies like the fragment guided selection of hydroxycoumarins, anthranilic acids and fenamic acids based on the types of aminoacids present in the inhibitor active site (Fritzson *et al.*, 2010). Another paper, describes Ascofuranone and Ascochlorin metabolites from a fungal strain show moderately good inhibitory action against hDHODH (Qi *et al.*, 2009). Preceding this work, we have used a database of 1000 natural compounds to evaluate good inhibitors of hDHODH using novel E-pharmacophore validation techniques (Swaminathan *et al.*, 2014). Here, our study deals with validating small molecules from traditionally used medicinal plant like *Cardiospermum halicacabum* as good inhibitors of hDHODH using new validation techniques.

* Corresponding Author
Email: saleena.m@ktr.srmuniv.ac.in

The validation is a multi course process where multiple statistical and repetitive tests are done to evaluate the confidence of the results obtained. The reproducible results are further crossdocked to evaluate the ligand using flexible hDHODH receptors from RCSB PDB. The optimal poses of the lead-like compounds were shortlisted. These poses are taken as the base structure for molecular dynamics to understand the behavior of the suggested lead-like phytocompounds against hDHODH in a simulated soluble environment.

The phytomolecules used in the study are the active ingredients of Indian medicinal plant *Cardiospermum halicacabum*. *C. halicacabum* has been used by many traditional medicinal systems as treatment of RA (Ganesan *et al.*, 2011). Pharmacognasy properties of *C. halicacabum* as a crude drug have been studied and reported earlier (Zalke *et al.*, 2013). *In vitro* studies have shown ethanolic crude extract of the leaf of *C. halicacabum* inhibit nitric oxide synthase and TNF- α (Sheeba and Asha, 2009).

Other studies have been used to prove its anti-inflammatory nature (Babu and Krishnakumari, 2005; Sadique *et al.*, 1987). Pharmacological validation of this plant revealed that the plant leaf compounds are safe and are not toxic up to 40g/kg in murine models (Pillai and Santhakumari, 1981). In other studies, the plant has been evaluated for its antimicrobial characteristics (Jeyadevi *et al.*, 2012; Raza, 2013) as well as vectoricidal effects for mosquito borne diseases (Govindarajan and Sivakumar, 2012). Computational biology is a way to evaluate a probable inhibitor in a faster and definite manner in a workstation before experimentally determining the same (Marrone *et al.*, 1997; Jorgensen, 2004). Virtual screening and molecular docking validation is considered to be a good way to evaluate the significance of the result. The commonly used method for validation is ROC curve (Vyas, 2008). Reproducing the same results in the experimental laboratory seems to be the biggest challenge in an *in silico* study (Kirchmair *et al.*, 2015). Thus, this study helps to examine the results with respect to different aspects such as ranking, robustness, significance and pose difference to come forward with the best poses of the top hits of compounds from *C. halicacabum*.

This study would provide a comprehensive base for in depth analysis and search for alternative compounds that inhibit hDHODH and to explore an alternative mechanism of action for *C. halicacabum* inhibitors against hDHODH. The study provides a summative proof of concept for the use of *C. halicacabum* as a treatment for RA.

MATERIALS AND METHODS

Collection of SDF formats of Phytomolecules of *C. halicacabum* through Literature

The compounds to be interrogated for their mechanism in Rheumatoid arthritis against the target human DHODH was collected through literature search (Veeramani *et al.*, 2012; Jeyadevi *et al.*, 2012).

Protein Preparation

The protein preparation wizard workflow of Schrodinger 9.2v for the 3D-structures of proteins was implemented by adding hydrogens, assigning missing atoms and bond types, removing water molecules, and a refinement by completing missing side chains. The protein then was energy minimized using OPLS force field (Madhavi Sastry *et al.*, 2013).

E-Pharmacophore Screening

Using the Advanced Search Tab in Schrodinger Maestro Suite, the E-pharmacophore was imported against the 19 compounds of *Cardiospermum halicacabum*. The screening resulted in finding matches that are similar to the ADHR features in the E-Pharmacophore.

Ligand Preparation

Ligands were prepared using Ligprep program of Maestro 9.2v (Madhavi Sastry *et al.*, 2013). Each ligand was assigned an appropriate bond order and minimized using OPLS 2005 force field and ionized at pH 7.

Receptor Grid Generation

The receptor grid file was generated using receptor grid generation option of Schrodinger 9.2v which represents physical properties of a volume of the receptor (specifically the active site) that are explored while performing docking (Madhavi Sastry *et al.*, 2013). The grid is specified based on the ligand binding site in the co-crystal ligand.

Validation Methods for Docking Programs

Docking methods were evaluated using specificity, reproducibility using different programs and Receiver operating curves (ROC) plot. Docking using the decoy set of compounds that are mixed with the test compounds of a specific target. The ranking is used to plot receiver operating characteristic curves. ROC curves plot the sensitivity (Se) of a given docking / scoring combination against the specificity (Sp). ROC curve is a plot of true-positive versus false-positive rates for all compounds. The area under the ROC (AU-ROC) curve is the probability of active compounds being ranked earlier than decoy compounds. A ROC curve that passes through the upper left corner (100% sensitivity and 100% specificity) has perfect discrimination (no overlap in the two distributions). Therefore, the closer the ROC curve is to the upper left corner, the higher the overall accuracy of the test (Triballeau *et al.*, 2005).

Molecular Docking Using Glide XP (Extra Precision)

Molecular docking was performed for the prepared ligands using glide extra precision (XP) of Maestro 9.2v. This algorithm recognizes good hydrophobic, hydrogen-bonding, and metal-ligation interactions, and punishes steric clashes. Glide Score is based on ChemScore (Eldridge *et al.*, 1997), but includes a steric-clash term, adds buried polar terms devised by

Schrödinger to penalize electrostatic mismatches, and has modifications to other terms.

Molecular Docking Using Ligand Fit

The LigandFit docking procedure in Accelrys discovery studio consists of: cavity detection for active site; and docking ligands to a selected site. Cavity detection, involves the use of a flood-filling algorithm. Its scoring function has been derived to predict binding affinities or to prioritize ligands relative to one another (Venkatachalam *et al.*, 2003).

Molecular Docking using iGEMDOCK

iGEMDOCK v2.1 - A Graphical Environment for Recognizing Pharmacological Interactions and Virtual Screening[VS] is a docking tool for short listing pro-interactions like hydrogen bonds, electrostatic interactions and Van der Waals forces between the ligand and the receptor.

The energy value calculated by iGEMDOCK is the sum of binding energy, binding site pharmacophores energy and penalty of suboptimal ligand (Hsu *et al.*, 2011).

BindingDB

It is a web-accessible database of experimental IC50 of protein considered to be drug-targets with small, drug-like molecules (Nicola *et al.*, 2012). BindingDB contains 1,116,226 binding data, for 6,974 protein targets and 478,572 small molecules. It is expected that the enhanced access to data provided by this resource will facilitate drug-discovery and the development of predictive computer models of binding.

Crossdocking

To study the specificity of the receptors and ligands, when the receptors are all the same protein but in different conformations can yield valuable information about the effects of induced fit upon binding (Totrov and Abagyan, 2008). Glide Cross Docking (XGLIDE) automates cross docking calculations with Glide. The flexibility of the receptor is taken into account for ligand specificity. It is a practical method where experimentally derived structures are used to improve docking predictions

Free Energy Calculation using Prime MM/GBSA

Molecular mechanics with generalized Born and surface area solvation (MM/GBSA) is a popular approach performed using Prime option of Schroedinger's Maestro suite. MM/GBSA quantifies the difference in energy between the free and the complex state of both the ligand and the protein after energy minimization. The energies of the complex were calculated using OPLS 2005 (optimized potentials for liquid simulations force field) and generalized Born and surface area solvation model. This program is used to predict the strain energy and free energy of binding for each ligand (Du *et al.*, 2011).

Dynamics Simulation

A solvated system was subjected to energy minimization using steepest descent algorithm. MD simulations on Desmond

first relax and minimize the receptor alone or the complex. Further MD simulations were carried out for desired period of time at a constant temperature of 300K and constant pressure of 1 atm with a time step of 2fs by means of Nose-Hoover thermostat method. Molecular dynamics is a complement experiment to explain a molecular mechanism as well as describe a system at equilibrium (Karplus and McCammon, 2002). The energy and trajectory atomic coordinates for all simulations were recorded at every 1.2 ps and 4.8 ps.

RESULTS AND DISCUSSION

Phytochemicals of *C. halicacabum*

Cardiospermum halicacabum is an uncultivated food crop, widely distributed in India and consumed as green leafy vegetable. The anti-inflammatory activity of ethanol extract of *C. halicacabum* leaves against carrageenan-induced rat paw edema has already been observed experimentally in rats (Huang *et al.*, 2011). The ethanol extract of the *C. halicacabum* suppresses the production of TNF-alpha and nitric oxide in human peripheral blood mononuclear cells (Venkatesh and Krishnakumari, 2009). Jeyadevi *et al.*, (2012) worked on the ethanolic extract for its chemical constituents using GC-MS and FT-IR. This yielded in peaks for flavonoids and phenolic acids. Further literature searching resulted in some other compounds like cardiospermin, Calycosin, Rutin, Protocatechaldehyde, Quebrachitol, Phloridzin, Prunin and Ferulic acid (Veeramani *et al.*, 2012; Jeyadevi *et al.*, 2012). In total 19 compounds were shortlisted from the ethanolic leaf extract of *C. halicacabum*. Most of the compounds were flavanoids or alkaloids.

E-Pharmacophore Screening

A validated E-Pharmacophore for hDHODH published in a previously done work (Swaminathan *et al.*, 2014) was used to screen the 19 compounds. E- Pharmacophore or structure based pharmacophore is a scaffold of important features and their distances that specify the efficiency of an inhibitor. The validated E-Pharmacophore had 4 features viz. an Acceptor, a Donor, a hydrophobic group and a Ring structure. All these features correspond to the hydrogen bonds and hydrophobics interactions obtained in the X-ray Crystallographic Structure of hDHODH (1D3H) (Swaminathan *et al.*, 2014). The essential features represented in the E-pharmacophore ADHR at specific distances was screened against the 19 compounds to identify hits that have inhibitor like features. Fitness score measures the degree of equivalence between the ligand conformer and the hypothesis (Sotriffer, 2011). The E-pharmacophore which was based on A771726 - a known potent inhibitor of hDHODH was screened against the 19 compounds of *C. halicacabum* to sought out structurally similar compounds in the leaf that have similar features like the inhibitor. The E-pharmacophore with essential hDHODH inhibitory features was further taken for database screening to identify the novel molecules. **Table 1** lists the 8 compounds that showed fitness score equal or greater than 1.The

highest score obtained was for coniferaldehyde with 1.78 while 7 other compounds were equal to or higher than 1.3 fitness score.

Table 1: Fitness Scores of compounds screened with E-Pharmacophore.

Compounds	Fitness score
Coniferaldehyde	1.78
Quercetin	1.71
Chryseriol	1.67
Prunin	1.65
Apigenin	1.62
Calycosin	1.53
Caftaric acid	1.52
Luteolin7O- Glucornide	1.32

Table 1 suggests that coniferaldehyde and quercetin along with Chryseriol have similar features at distances matching with that of the pharmacophore hypothesis. All the hits were superimposed on the E-pharmacophore to manually inspect the degree of similarity (**Figure 1**). Thus compounds that are top hits in pharmacophore screening may dock with higher score. It is also accomplished that all of the 8 compounds obtained have features that match at various degrees to the known inhibitor of human DHODH.

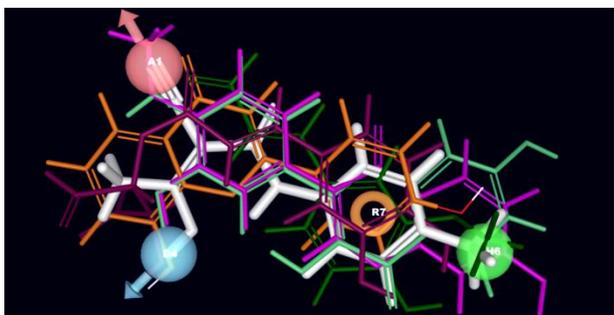


Fig. 1: Superimposition of the different hits after screening with the E-pharmacophore (White), Coniferaldehyde(green), Quercetin(cyan), apigenin (orange), calycosin(magenta) and caftaric acid(purple).

Glide Docking

Although only fewer hits were screened in the previous step, all 19 phytochemicals were docked to identify the binding mode of the active phytoconstituents with the inhibitor active site of human DHODH receptor. The glide XP G-score ranged from -11.28 Kcal/mol to -8.22 Kcal/mol. The hit compounds which possess high glide score compared with the known inhibitor A771726 (Glide XP G-score= -8.15 Kcal/mol) was considered as good hits (**Table 2**).

Out of 19 compounds, only ten compounds showed XP G-score greater than A771726. The XP G-score is generated using scoring functions that consider the bonded and non-bonded interactions as well as penalties for rotatable bonds and polar group interactions in active site. These hit compounds were examined for their bonded and non bonded interactions with human DHODH. In an earlier study done to find the most effective inhibitor against human DHODH, it was discussed that an ideal human DHODH inhibitor must possess Van der Waals, hydrophobic and negatively charged interactions with the receptor.

The binding mode of the identified hits was inferred by studying the essential bonded and non-bonded interactions. Our identified hits show good electrostatic (XP electro) and hydrophobic interactions (XP PhobEn) with the receptor DHODH. A comparison of docking poses for known and identified hits suggests that the hits show almost similar binding mode with the active site residues of DHODH (Pro52, Ala55, Thr63, Leu67, Arg136 and Tyr356).

Table 2: Docking results for the phytochemicals against human DHODH using GlideXP.

S.No	Compound Name	Glide XP Score (Kcal/mol)	Glide Energy (Kcal/mol)	Interactions
1.	Caftaric Acid	-11.28	-45.16	Ser215,Asn284, Asn212, Lys100,Ser120 Tyr356, Pro52
2.	Luteolin7O-glucornide	-10.85	-55.36	Tyr 38
3.	Phloridzin	-10.44	-46.43	Pro52
4.	Prunin	-10.10	-37.03	Pro52(2)
5.	Chlorogenic acid	-9.84	-50.02	-
6.	Chryseriol	-9.18	-39.47	Leu359
7.	Calycosin	-9.05	-37.92	Ala55
8.	Apigenin	-8.69	-37.61	Pro52
9.	Apigenin7O-glucornide	-8.22	-57.87	-
10.	A771726	-8.15	-41.02	Tyr356,Arg136

Binding Mode of Caftaric Acid

Caftaric acid, which was one of the hits in the E-pharmacophore screening, had the highest Glide XP Gscore of -11.22 Kcal/mol among other hits. This compound revealed a network of hydrogen bonds with receptor residues like Ser215, Asn284, Asn212, Lys100 and Ser120. This compound donates a Hydrogen bond with side chain of Asn212(1.73 Å) and accepts 4 hydrogen bonds from the side chains of Ser215, Asn284, Lys100 and Ser120 at distances 2.02Å, 1.92Å, 1.74Å, 1.85Å respectively (**Figure 2**). However the binding mode is not in the same pocket as that of the known inhibitor site but farther away in a deeper pocket near substrate binding site of FMN. Thus, compared to A771726, caftaric acid binds in a region away from the known binding site and thus cannot be considered as a good hit among all the hits obtained as compared to the known inhibitor A771726. It has rich hydrophobic interactions with Tyr147, Val134, Tyr356 and Val143.

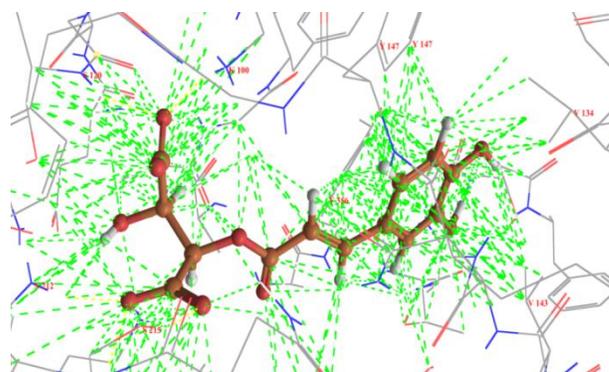


Fig. 2: Binding mode of Caftaric acid.

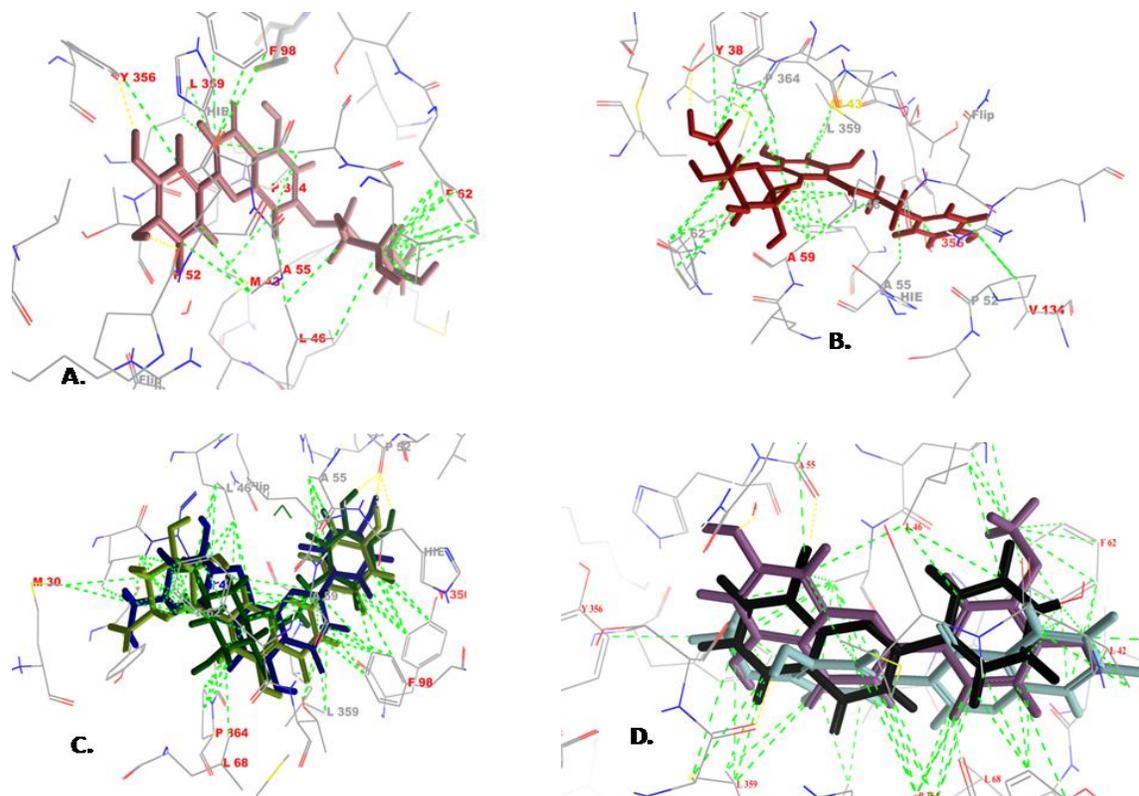


Fig. 3: Binding modes of A. Luteolin 7O-glucuronide B. Phloridizin C. Prunin (green) Chlorogenic acid (black) and Apigenin 7O-glucuronide (blue) D. Chryseriol (pink), Calycosin (purple) and Apigenin (black)

Binding Mode of Luteolin-7O Glucuronide

Rhamnazin, a flavonol, scores -10.70 Kcal/Mol and donates a hydrogen bond to a main chain of Pro52 backbone at the carboxyl oxygen (1.71 Å) and donates another hydrogen to the side chain of Tyr356(2.02Å). Its XP vdW and XP PhobEn are -0.69 and -0.17 respectively. Its hydrophobic score is much better than the other molecules compared. It makes hydrophobic interactions with Met43, Leu46, Pro52, Ala55, Phe62, Phe98, Pro364, Tyr356 and Leu359. The binding mode is quite similar to the binding mode of A771726 with Tyr356 (2.91 Å). The hydrophobic interactions of this molecule contribute to the high Gscore obtained and indicates a possible effective inhibitor activity of the compound against human DHODH (**Figure 3A**).

Binding Mode of Phloridizin

Phloridizin is a flavonol that binds with a Gscore of -10.44Kcal/mol and forms a hydrogen bond with Tyr38 (1.88 Å). The side chain of Tyr38 accepts hydrogen atoms from the Hydroxyl group present in the dihydroxyphenyl ring of phloridizin. The electrostatic interaction contributes to increase in the XP Electro contribution to -0.25. The XP Vdw contribution is about -43.07 with hydrophobic bonds formed with Tyr38, Leu42, Met43, Leu46, Pro52, Ala55, Ala59, Leu68, Val134, Leu359 and Pro364. Tyr38 is in the same pocket as that of Tyr356 thus the binding mode is comparable with A771726 (**Figure 3B**).

Binding Mode of Prunin, Chlorogenic Acid and Apigenin 7O - Glucuronide

Prunin, chlorogenic acid and apigenin 7Oglucuronide all interact with the amino acid Pro52 forming bonds at an average distance of hydroxyl group at 2.1Å. Chlorogenic acid forms two bonds with Pro52. The high XPGscore for prunin is accounted by the high Vdw interaction score of -32.96 Kcal/mol which is comparable with that of A771726 having a score of -31.94 Kcal/mol. Apigenin 7Oglucuronide shows similar XP PhobEn of -2.82 with A771726 having XPPhoben score of -2.00. The Hydrophobic interactions with Met43, Leu46, Ala55, Ala59, Phe62 and Leu67 and Tyr356 and Leu364 were common for the three molecules compared (**Figure 3C**). Chlorogenic acid had high XP Phoben and XP vdw compared apigenin 7O glucuronide but lower than Prunin.

Binding Mode of Chryseriol, Calycosin and Apigenin

Among the three, Chryseriol has the highest XP PhobEn contribution compared to the others. They had a glide XPGscore of -9.18, -9.05 and -8.69 Kcal/mol. The amino acid interactions forming Hydrogen bonds were with Ala55 and Leu359 (**Figure 3D**). The hydrophobic interactions were with the amino acids Met43, Leu46, Ala55, Ala59, Phe62 and Leu67 and Tyr356 and Leu364. The molecules are consecutive in the list of glide XP results and only differ majorly in their glide evdw score of -37.37, -33.33 and -34.43 respectively. Based on the binding mode

analysis it is clear that all the hits obtained are comparable to the known inhibitor and its significance is maintained by binding in an equivalent manner as the known inhibitor molecule. The exception in this is Caftaric acid which has a good glide score and energy but the Binding pocket is a little deeper than the inhibitor binding site. To evaluate if caftaric acid is a significant hit, docking experiment is validated by using three validation methods.

Validation of the Docking Results

The docking results significance can be validated by many methods. The validation aims to corroborate the robustness of the results with its reliability. The results will throw light on the factors that affect a docking experiment, like the docking algorithm, the use of large data set of decoys and performance of the docking tool with known and unknown active molecules.

ROC Plot Analysis

Receiver operator Characteristic [ROC] curve is done to justify the sensitivity and the specificity of the receptor towards the phytochemicals. Docking program was validated using ROC which describes the ability of docking method to avoid false positives and false negatives. The docking experiment is modified to accommodate the 19 compounds of *C. halicacabum* and 300 compounds from the Asinex decoy set in the ratio 1:15. The decoys have been used traditionally to evaluate the sensitivity of a pharmacophore while in other studies have been used for evaluating the specificity of a receptor. Enrichment was calculated based on how well the candidate or annotated compounds were retrieved from the decoy database. Here the ROC curve obtained had an Area under Curve (AUC) value of 0.72 which is significant according to literature survey (Figure 4). 79% of the actives compounds (19 compounds) were retrieved as hits. The top four hits of Glide XP were in correlation to the top four hits of the decoy set infused actives docking.

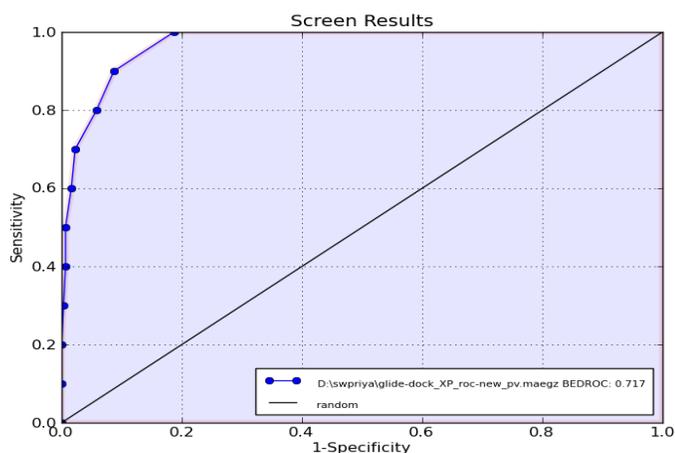


Fig. 4: Receiver Operator characteristic curve against decoys

Docking with Several Different Algorithms

The docking algorithm plays a humongous role in retrieving the true positives as results. Different algorithms were

used to check the robustness of the result obtained. Two docking algorithms were used for docking the 19 phytochemicals of *C. halicacabum* with the receptor 1D3H.

Docking In Accelrys Ligandfit Tool

The docking algorithm of Ligandfit is different from GlideXP as it searches for shape complementarities of the ligand with the receptor rather than hierarchical filtering of ligand poses in the receptor pocket. This amounts to searching of ligand pose in all bumps or openings in the receptor structure. Ligandfit was able to define the receptor to have 3 binding pockets. When the 19 ligands along with A771726 were probed in 3 different pockets, good dockscore and PLP score were obtained for the different ligands only in the inhibitor binding site.

Table 3 shows 7 hits having better dockscore compared to A771726. The dockscore ranges from 73.01 to 55.78 Kcal/mol.

Table 3: Docking results for the phytochemicals against human DHODH using Accelrys Ligandfit.

Compound Name	Dockscore Kcal/mol	PLP1 Kcal/mol	PMF Kcal/mol	Interactions
luteolin7O-glucornide	73.01	-121.77	-122.38	Arg136, Tyr38, Pro52, Gln47
apigenin7O-glucornide	70.90	-114.15	-116.5	Pro364, Tyr38
prunin	67.48	-117.48	-100.99	Arg136
chlorogenic acid	65.06	-99.3	-80.57	Arg136, Pro 52, Gln47
chryseriol	63.01	-102.1	-72.55	Arg136, Tyr356, Pro 52, -
phloridzin	59.35	-104.28	-88.9	Thr63, Met43
caftaric acid	57.52	-88.22	-60.35	Arg136
A771726	55.78	-102.7	-67.4	Tyr356, Arg136

Luteolin 7O- glucornide was the top hit here compared to caftaric acid in Glide XP. The type of interactions made by Luteolin 7O glucornide is the similar with Pro52. The bond distance is comparable. All the other hits including caftaric acid bound in the same pocket as that of the inhibitor Leflunomide. Figure 5 gives the binding pose of the top hit.

The remaining two predicted binding sites gave “no poses docked”. This means nil results were obtained.

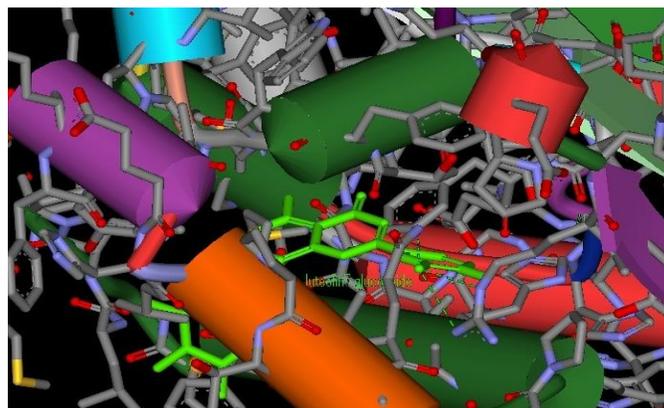


Fig. 5: Docked pose of Luteolin 7O-glucornide (light green) with hDHODH as predicted by Accelrys Ligand Fit tool.

Docking with the iGEMDOCK Tool

iGEMDOCK tool uses evolutionary steps of genetic algorithm to find docked poses of ligands. The poses generated are survivors of goodness of fit score. The docking results of the 19 phytocompounds with the human DHODH receptor gave 6 hits with Luteolin 7O-glucornide as the best hit followed by apigenin 7O glucornide. The hits generated were all better energy value compared to the known inhibitor A771726. The evolutionary based algorithms tends to select those hits that are better among a population of hits or poses that have better energy than the earlier hits obtained. This iterative process only ceases when no better pose with better energy can be generated.

Table 4: Docking results for the phytomolecules against human DHODH using iGEMDOCK tool.

Compound Name	Energy Kcal/mol	Hydrogen bond Interactions	Electrostatic Interactions	Vaanderwaals Interactions
Luteolin7O-glucornide	-153.22	Arg136,Tyr38,Pro52, Gln47, His56,Thr63, Tyr356	-	Ala59, Met43, His56, Ala55, Phe62, Phe98, Leu359, Thr360, Pro364
Apigenin7O-glucornide	-147.00	Arg136,Tyr38,Pro52,Gln47,His56, Tyr356	Arg136	Pro364, Tyr38, Pro52, Ala55, His56, Ala59, Phe62, Phe98, Leu359,Thr360
Phloridzin	-141.52	Arg136,Pro52, Gln47, Ala59, Leu68,	-	Pro52, Ala55, His56, Ala59, Phe62, Arg136, Tyr356,Leu359 Thr360 Pro364
prunin	-141.10	Arg136,Thr63, Tyr38	-	Ala59, Met43, His56, Ala55, Phe62, Phe98, Leu359,Thr360 Pro364, Tyr38, Arg136
chlorogenic acid	-138.50	Arg136,Tyr147,Thr63, His56, Tyr356	-	Arg136,Tyr356,Pro52,Ala55,, His56,Phe98, Val134,Leu359, Thr360
caftaric acid	-124.09	Gln47,Thr63,Arg136, Tyr356, Pro364	Arg136	Gln47,Pro52, Ala55,His56,Ala59, Phe98, Arg136,Tyr356 Leu359,Thr360
A771726	-123.8	Tyr356, Arg136	-	Gln47,Pro52, Ala55,His56,Ala59,Phe98,Arg136,Tyr356, Leu359,Thr360, Pro364

Table 4 summarizes the iGEMDOCK result in such a way that it gives details of the hits in the format of Energy and the nonbonded interactions like hydrogen bonds, Van der Waals interactions and electrostatic interactions. The energy score ranged from -153.22 to -123.8 Kcal/mol. Luteolin 7O-glucornide and

Apigenin 7O- glucornide were the top two hits. All the top hits had similar interaction in the inhibitor site. **Figure 6** depicts the binding pose for Luteolin 7O-glucornide as predicted by iGEMDOCK. It shows Hydrogen bonds as dashed yellow lines with Arg136, Tyr38, Pro52, Gln47, His56 and Tyr356.

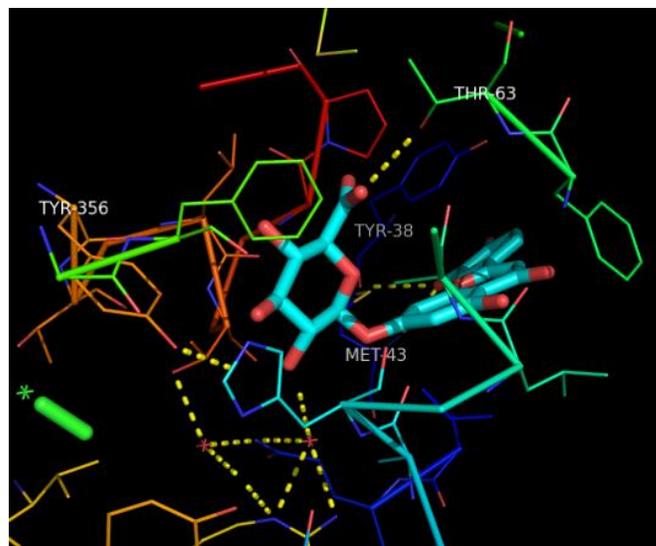


Fig. 6: Docked pose of top hit (luteolin7O–glucornide) as predicted by iGEMDOCK.

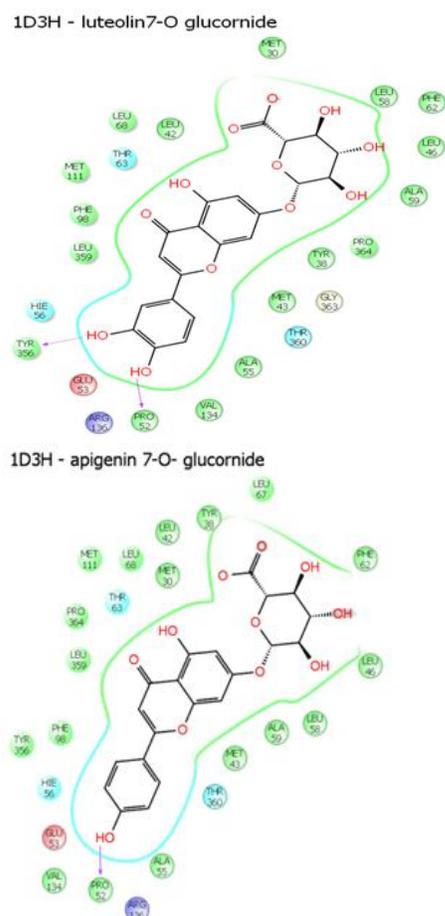
Validation using Curated BindingDB

BindingDB is a database that contains binding affinity information of different receptors for different ligands in different organisms. The database was curated for small molecules that inhibit human DHODH specifically and which neither binds with other species DHODH receptor nor with other protein drug target receptors. Such potent, crosschecked and specific small molecules were collected and made into a phase database file as curated BindingDB for human DHODH. A total of 229 molecules were found that were very specific to human DHODH with high IC50 value. These inhibitors along with the nineteen phytomolecules were docked with human DHODH to figure out where the phytomolecules rank themselves compared to known, potent, specific inhibitors. 333 poses were generated where caftaric acid and Luteolin 7O-glucornide and appeared among the top 25 of the hits. A771726 among other known inhibitors was ranked 155 with a XP GScore of -8.32 Kcal/mol. Caftaric acid and Luteolin 7O glucornide had XP Gscore of -11.28 and -10.85 Kcal/mol which is significantly higher. However the binding pocket differed for caftaric acid in a deeper tunnel farther away from the inhibitor binding site. Other compounds which were ranked well than A771726 were phloridzin, prunin, chlorogenic acid, chryseriol, calycosin and apigenin. Their XP Gscore and binding site were analogous with A771726.

The **Table 5** depicts the hits when sorted with glide energy. Table depicts the known compounds with its IC50 value while the *C. halicacabum* compounds do not have a known IC50 value against the target.

Table 5: Glide XP score of Binding DB based known inhibitors against hDHODH.

S. No	TITLE	ENZYMOLOGIC IC50 (nM)	GLIDE ENERGY Kcal/mol
1	apigenin7O- glucornide	-	-57.87
2	luteolin7O- glucornide	-	-55.36
3	CHEMBL370008	2000	-51.65
4	CHEMBL2177115	1270	-51.16
5	CHEMBL154336	370	-50.71
6	US8536165	190	-50.09
7	chlorogenic acid	-	-50.02
8	Phloridzin	-	-49.66
9	CHEMBL155548	290	-49.43
10	CHEMBL1076869	3300	-49.38
11	CHEMBL358435	8400	-49.35
12	CHEMBL154908	2900	-49.20
13	CHEMBL193365	11	-49.11
14	CHEMBL1929439	27	-48.95
15	Biphenyl-4-ylcarbamoyl Thiophene Analog 5	44	-48.61
16	Biphenyl-4-ylcarbamoyl Thiophene Analog 8	1	-48.42

**Fig. 7:** Ligand Interaction Diagram for Luteolin 7O- glucornide and Apigenin 7O-glucornide.

Glide energy is the energy associated with binding of ligand with receptor. When the results were sorted according to Glide energy it was seen that Apigenin 7O- glucornide and Luteolin 7O- glucornide were the best hits. More negative energy indicates better binding. Thus the results suggest that compared to

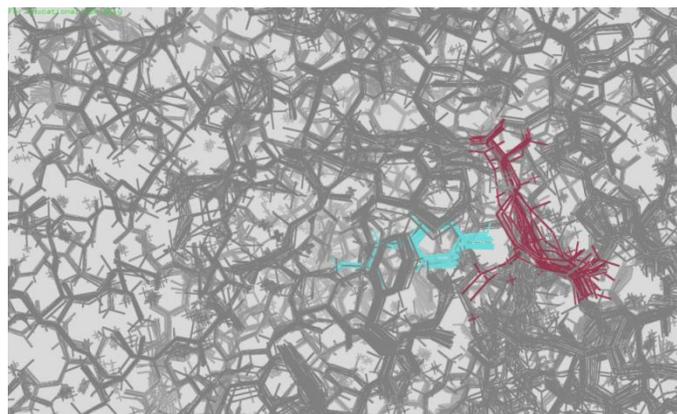
already known inhibitors the results obtained by docking with *C. halicacabum* phytomolecules were very significant. **Figure 7** gives the Ligand Interaction Diagram description of Luteolin 7O-glucornide and Apigenin 7O-glucornide when docked with several known BindingDB inhibitors. It was seen that the two structures are chemically similar to each other except for a hydroxyl group missing in apigenin 7 O-glucornide. It was also observed that their binding mode is similar with hydrogen bonds at Pro52 of hDHODH.

Cross Docking

Regular molecular docking considers the conformation of the target receptor to be rigid and the ligand to be flexible. This rigidity can account for the false positivity of the docking results. Thus a method called CROSS docking allows docking of all the different conformations of DHODH present in PDB with the 19 compounds. Cross docking tends to retrieve significant top hits with different conformations of the receptor to ultimately eliminate false positives. It's a class of flexible docking with receptor and ligand flexibility that justifies the docking poses of the ligand with different conformations of the receptor.

There were 22 different other receptors of DHODH for humans at the time of doing the experiment having a resolution ranging from 1.24Å to 3.00 Å.

The **Figure 8** gives a glimpse of all the 22 PDB structures of human DHODH superimposed on each other. The raspberry colored residue is ARG136 and cyan coloured residue is TYR356- part of the inhibitor binding site.

**Fig. 8:** Superimposed image of 22 PDB structures of hDHODH.

The cross docking results **Table 6** depicts the best 3 hits of each receptor of human DHODH are displayed as the result table of cross docking. Luteolin 7O-glucornide had the highest XP G-score of -15.913 Kcal/mol for the PDB structure 2BOM. For all the 22 cross docked results A771726 was lower than the top three hits. It was seen that all the resolutions from lowest to highest gave the same compounds as hits as from regular docking. Each hit was counted for the number of times it occurred as the first, second or the third hit for the different PDB id receptors.

Table 6: Cross Docking score for 22 different DHODH based PDB Structures.

PBD ID	Compound Name	XP G-SCORE	GLIDE ENERGY	Interactions
		Kcal/mol	Kcal/mol	
2BOM	Luteolin 7O-glucornide	-15.978	-57.419	TYR356,ARG136,THR360
	Apigenin 7O-glucornide	-12.773	-45.669	PRO52,GLN47, ARG136,THR360
	Prunin	-12.624	-62.943	PRO52,ARG136,THR360
2PRH	Luteolin 7O-glucornide	-15.821	-66.246	PHE62,LEU67, ARG136,THR360
	Prunin	-12.560	-52.889	TYR356,ARG136,VAL367
	Apigenin 7O-glucornide	-12.254	-56.497	TYR38,ARG136
1D3G	Luteolin 7O-glucornide	-12.647	-60.206	TYR38,ARG136(2)
	Prunin	-11.365	-51.671	ARG136,LEU(67)
	Chlorogenic acid	-11.332	-54.810	PRO52,TYR38
2BXV	Apigenin 7O-glucornide	-11.209	-59.486	MET30,LEU50,ARG136
	Chlorogenic acid	-10.766	-54.689	GLN47,TYR38,LEU67
	Luteolin 7O-glucornide	-10.645	-54.488	LEU42,ALA55
2PRL	Luteolin 7O-glucornide	-14.804	-64.290	PRO52, LUE67,TYR356,ARG136
	Apigenin 7 O-glucornide	-13.595	-46.383	PRO52,GLN47,THR360,LEU359
	Prunin	-12.592	-47.084	PRO52,ARG136,
2WV8	Luteolin 7 O-glucornide	-13.165	-60.261	PRO52, TYR356,ARG136
	Apigenin 7 O-glucornide	-12.543	-52.731	GLN47,PRO52,ARG136
	Prunin	-10.403	-47.985	PRO52,GLN47
2FPT	Luteolin 7 O-glucornide	-12.908	-51.039	GLN47,ARG136
	Apigenin	-10.845	-50.559	ARG136(2)
	Chlorogenic acid	-10.811	-52.972	ARG136,PRO52
2FPV	Luteolin 7 O-glucornide	-13.666	-66.036	TYR356,ARG136,PRO52
	Prunin	-11.823	-5.597	ARG136,PRO52,TYR38
	Chlorogenic acid	-10.509	-53.167	ARG136,PRO52,TYR38
3FJ6	Luteolin 7 O-glucornide	-13.883	-55.947	PRO52, LUE67,ARG136
	Apigenin 7 O-glucornide	-12.155	-53.138	PRO52,ARG136,THR360
	Chlorogenic acid	-11.066	-57.702	TYR38,PRO52,ARG136
2FPY	Chlorogenic acid	-12.051	-55.551	TYR38,ARG136
	Luteolin 7 O-glucornide	-11.595	-59.333	TYR38,GLN47,ARG136
	Caftaric acid	-11.138	-56.611	LEU67,TYR360
2FQI	Luteolin 7 O-glucornide	-12.565	-58.606	TYR38,GLN47,ARG136
	Chlorogenic acid	-11.566	-58.305	TYR38,ARG136,GLN47,PRO52,
	Caftaric acid	-11.472	-56.901	LEU67,TYR38,ARG136,GLN47,PRO52
3F1Q	Chlorogenic acid	-11.157	-49.828	PRO52
	Luteolin 7 O-glucornide	-10.290	-50.543	LEU42
	Phloridzin	-9.716	-48.027	PRO52
3FJL	Luteolin 7 O-glucornide	-10.484	-42.536	PRO52,ARG136
	Phloridzin	-10.196	-39.048	LUE42,THR63,TYR356
	Apigenin 7 O-glucornide	-9.950	-54.281	PRO52,LEU42
3GOX	Apigenin 7 O-glucornide	-12.469	-53.739	PRO52,ARG136(2),GLN47
	Luteolin 7 O-glucornide	-11.969	-60.624	ARG136,TYR38
	Prunin	-11.029	-53.645	GLN47,PRO52, ARG136
3GOU	Luteolin 7 O-glucornide	-11.746	-62.255	PRO52,ARG136,GLN47,THR360
	Apigenin 7 O-glucornide	-10.095	-55.482	ARG136,TYR38
	Phloridzin	-9.601	-47.349	ARG136,TYR38
3U2O	Prunin	-12.715	-54.631	TYR356,PRO52(2),ARG136(2),THR360
	Luteolin 7 O-glucornide	-11.736	-61.162	ARG136(3),PRO52
	Apigenin 7 O-glucornide	-10.923	-59.133	LEU67,ARG136
3W7R	Luteolin 7 O-glucornide	-14.443	-65.248	ARG136(2),PRO52,TYR356,LEU67
	Apigenin 7 O-glucornide	-14.078	-59.445	GLN47,ARG136(2)
	Caftaric acid	-11.770	-57.660	LEU67(2),ARG136,THR360,PRO52
3ZWS	Caftaric acid	-10.614	-54.308	LEU67,TYR356,PRO52,THR63
	Chlorogenic acid	-10.612	-46.071	ARG136,GLN47
	Phloridzin	-9.249	-42.920	LEU67,HIS41(2)
4JGD	Phloridzin	-11.510	-53.487	TYR38,THR63,LEU359
	Luteolin 7 O-glucornide	-11.210	-55.325	ARG136,TYR38
	Apigenin 7 O-glucornide	-10.547	-53.397	LEU67,PRO52
4JS3	Luteolin 7 O-glucornide	-12.228	-54.872	LEU68(2),TYR38
	Prunin	-10.963	-46.263	TYR38,LEU68
	Apigenin 7 O-glucornide	-10.827	-51.582	PHE62,LEU68(2)
4JTT	Luteolin 7 O-glucornide	-11.446	-53.933	LEU68(2),PHE62
	Phloridzin	-10.682	-45.420	LEU359,PHE62,LEU68,TYR38
	Apigenin 7 O-glucornide	-10.639	-49.430	LEU68(2),TYR38
4JTU	Luteolin 7 O-glucornide	-14.078	-67.862	ARG136(2),TYR356,PRO52
	Phloridzin	-11.460	-48.299	LEU67,ARG136
	Apigenin 7 O-glucornide	-11.013	-58.578	LEU67,ARG136

The **Table 7** below summarizes the cross docking results. It was seen that out of 22 cross docking results Luteolin 7O-glucornide was obtained as the best hit majority of the time (15 times). This was followed by Apigenin 7O- glucornide that was obtained as the second or the third hit majority of the times. These results indicate that Luteolin 7O- glucornide binds consistently with various conformations of human DHODH. The other compounds that were significantly bound to flexible hDHODH were prunin, Phloridzin, apigenin and caftaric acid. Then again caftaric acid did not bind in the same pocket as the other phytomolecules.

Table 7: Frequency of Hits in Cross Docking.

COMPOUND	As 1st HIT	As 2nd HIT	As 3rd HIT
Luteolin 7 O-Glucornide	15	4	1
Apigenin 7 O-Glucornide	2	6	7
Prunin	1	3	4
Phloridzin	1	3	3
Caftaric Acid	2	2	3
Chlorogenic Acid	1	3	5
Apigenin	0	1	0

Cross docking results emphasized on Luteolin 7O-glucornide and Apigenin 7O- glucornide to be significant along with all the previous validation methods used.

MM/GBSA

Any molecular docking experiment should ultimately be extrapolated to its dynamic behavior using molecular simulation. Thus the molecules to be simulated are decided upon by performing free binding energy calculations. MM/GBSA helps to hierarchically list the top best binding molecules of *C. halicacabum*. MM/GBSA predicted Apigenin 7O- glucornide followed by Luteolin 7O- glucornide as compounds having good binding energy compared to other phytochemicals of *C. halicacabum*. **Table 8** gives the Binding free energy and its individual contributions along with the results obtained in Glide XP. The compounds are listed in the hierarchy as obtained from Glide XP. Only two compounds like Apigenin 7O-glucornide) and Luteolin 7O-glucornide (-79.676 Kcal/mol) showed better binding free energy than A771726 while caftaric acid had the lowest dG Bind score (-35.346 Kcal/mol). So out of the 6 top results obtained in Glide and previously employed methods only two compounds had significant dG Bind in **Table 8**. The lowest score is that of caftaric acid.

MM/GBSA results indicated that the contribution of Vanderwals (vdW) was higher in the top two protein ligand complexes (Apigenin 7O-glucornide and Luteolin 7O- glucornide) compared to the known ligand A771726 (Leflunomide). The vdW affinities were (-60.24, -51.17 and -34.54 Kcal/mol) from former to latter. On comparison of all the attributes predicted by MM/GBSA it was clearly noted that Lipophilic (Lipo) contribution is also a main component for binding affinities with Apigenin 7O-glucornide and Luteolin 7O-glucornide having lipophilic binding energy as -45.16 and -46.98 Kcal/mol resp. compared to -31.02 Kcal/mol for A771726. The other low scoring compounds like

caftaric acid showed higher Columbic and solvation energies (-145.9 and -164.23 Kcal/mol respectively). These comparisons prove that good inhibitors show better vdW and Lipophilic contributions compared to non inhibitors. In other words better the lipophilic and Van der Waals interactions, better the binding energy of the inhibitor. Thus based on the MM/GBSA results Apigenin 7 O-glucornide and Luteolin 7O-glucornide were further taken for dynamics studies and compared with standard simulation of A771726 and Apo form of human DHODH protein.

Table 8: MM/GBSA results for binding free energy of ligand- DHODH receptor complex.

Compound name	Glide score (Kcal/mol)	DG bind (Kcal/mol)	DG_Coul	DG_Vdw	DG_Lipo	SolvGB
Caftaric Acid	-11.286	-35.346	-145.929	-32.969	-21.167	164.233
Luteolin 7 O-Glucornide	-10.842	-79.676	-39.353	-51.612	-46.986	50.549
Prunin	-10.109	-73.347	-10.935	-52.610	-51.170	17.005
Chlorogenic acid	-9.842	-73.314	-40.084	-53.917	-43.140	57.615
Chryseriol	-9.186	-67.408	-11.885	-38.950	-35.922	13.274
Apigenin 7 O-Glucornide	-8.213	-80.659	-31.600	-60.246	-45.162	46.341
A771726	-8.157	-75.449	-16.163	-34.549	-31.026	4.56
Cardiospermin	-8.121	-60.776	-10.441	-31.338	-38.655	13.398

Molecular Dynamics

Four complexes were taken for molecular simulation to evaluate their binding affinities with the human DHODH receptor. The apo form of hDHODH, hDHODH complexed with A771726, hDHODH complexed with Luteolin 7O-glucornide and hDHODH complexed with Apigenin 7O-glucornide was simulated for 10 ns and their RMSD was compared with that of A771726 and apo to check whether the hit compounds show a stable behavior with the receptor which is comparable with A771726 behavior and better than the apo receptor's behavior in a simulated environment of water molecules. The complexed proteins and their starting poses were taken from the cross docking files. Root mean square deviation [RMSD] is best means to measure the native behavior of a molecule in a simulation. **Figure 9** shows the comparative plot of RMSD for the 4 receptor complexes. From the graph it was seen that among the 4 simulations, apo receptor showed a higher RMSD than known inhibitor, A771726 by 0.22 Å while all the other simulations showed an RMSD closer to 1.5 Å which is widely acceptable. This indicates that all the 4 simulations were stable after the 10ns MD simulation.

Table 9: Average RMSD of Receptor complexed with different ligands.

hDHODH complexed with	Average RMSD
Apo protein	1.59
Luteolin 7O-glucornide	1.53
Apigenin 7O-glucornide	1.52
A771726	1.51

The lowest RMSD was exhibited by Luteolin 7O-glucornide complex at 1.46 Å. **Table 9** gives the average RMSD values of the receptor with the ligand throughout 10ns MD

simulation., it seen that Luteolin 7Oglucuronide (1.53\AA) and Apigenin 7O-glucuronide (1.52\AA) are quite steady throughout the 10ns simulation from **Table 9**. Lower RMSD indicates good dynamics. To understand the stability factor in a better manner the number of H-bonds formed between the hit compounds and receptor during 10ns simulation time was analyzed. The H-bonds analysis (**Figure 10**) showed that **Luteolin 7O-glucuronide** maintains on an average 2 hydrogen bond with residues like Tyr 356 and additionally interacts with Arg136, Ala55 throughout the whole simulation. Apigenin 7O-glucuronide maintains one hydrogen bond with the inhibitor site residues, except from 4000ps to 6000ps where its forms additional 3 H bonds on an average. In this period some water molecules come in closer contact with the inhibitor site thus leading to additional bonds. Apo was not analyzed for H bonds at the inhibitor site. The A771726 receptor complex started with 2 H bonds initially but is initially begun with 2 H-bonds with Arg136 but all the H bond disappeared by the first 1000ps itself which later re-establishes at 5000ps to form partially stable bond with Tyr147 which again disappears over time.

These H-bond interactions clearly indicate that the protein-ligand complexes of the hit compounds are stable and maintained with similar binding mode in the inhibitor site of human DHODH in the MD simulation studies. **Figure 11** shows the RMSF plot of the residues present in the inhibitor binding site. The root mean square fluctuation was the least for Apigenin 7O-glucuronide with 0.89\AA followed by 1.22\AA for Luteolin 7O-glucuronide complex. Apo protein and A771726 complex receptor had 1.33\AA fluctuations. At the 7th ns simulation time all the complexes fluctuate to 6.5\AA because of the water molecules intervention.

Figure 12 plots the ROG behavior of the simulated complexes. From the diagram it is quite clear that the radius of Gyration is predictably highest for apo DHODH (20\AA) and the lowest for A771726 complex (16\AA). The hits compound complex had an average ROG value of 19\AA at the end of 10ns. From MD studies we infer that all hit compounds occupy the active site of human DHODH as known inhibitor, A771726 indicating that the hit compounds have similar bonded and non-bonded interactions.

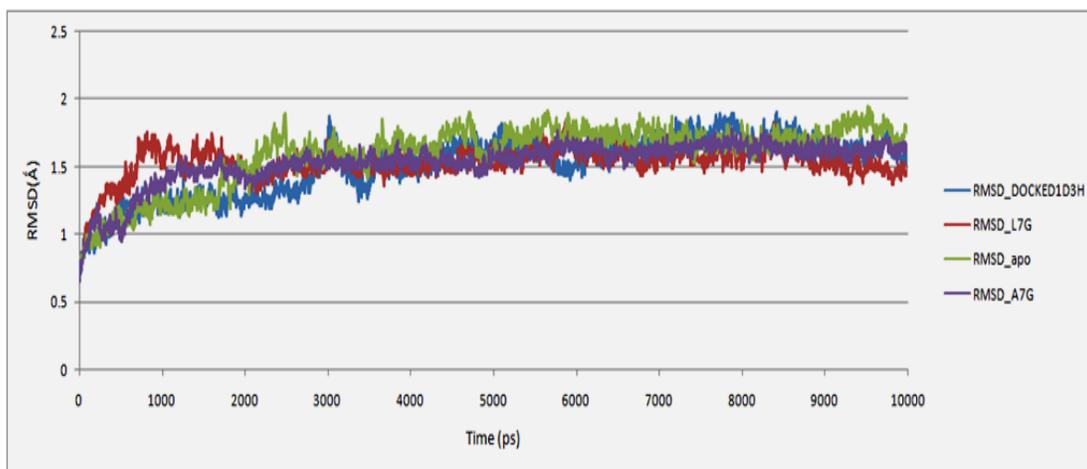


Fig. 9: RMSD plot for Backbone of Human DHODH docked with different ligands for 10ns.

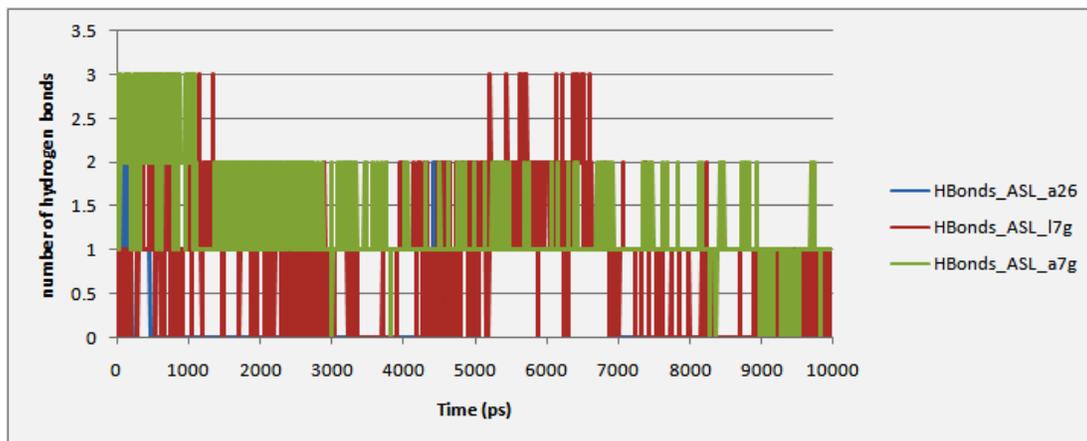


Fig. 10: Number of Hydrogen bond for (Arg136, Tyr356) of Human DHODH docked with different ligands for 10ns.

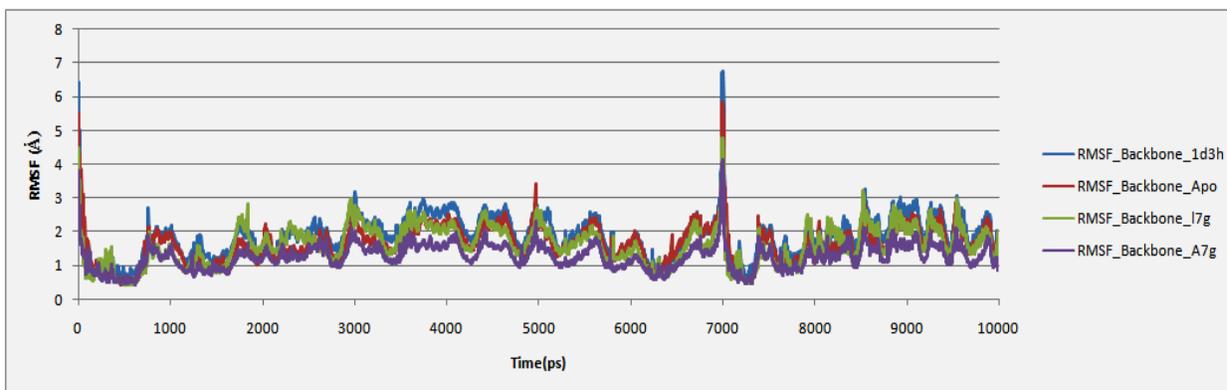


Fig. 11: RMSF for (Arg136, Tyr356) of hDHODH docked with different ligands for 10ns.

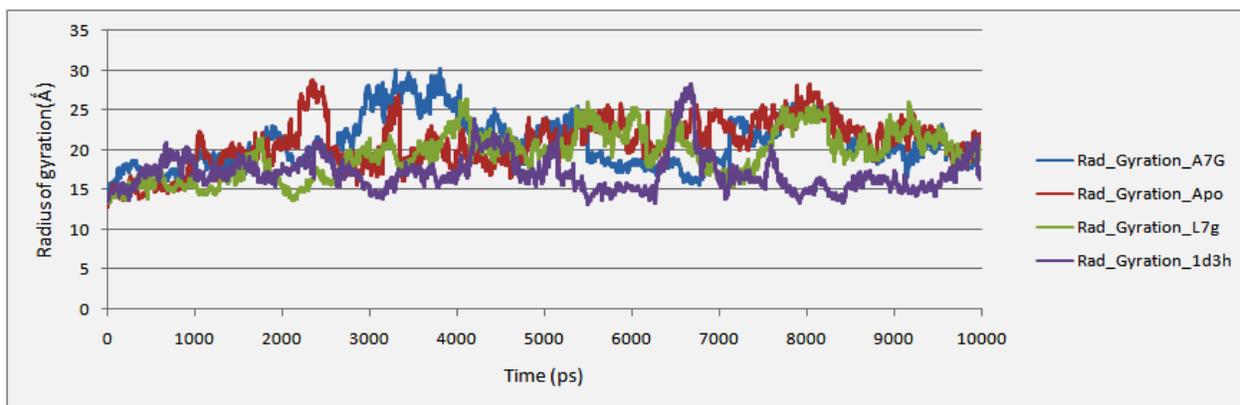


Fig. 12: ROG plot for Backbone of hDHODH docked with different ligands for 10ns.

DISCUSSION

C. halicacabum is a traditional medicinal plant used mainly for arthritis pain. It is widely known for its anti-inflammatory properties in inflammatory diseases. The Leaf compounds were corroborated as influencing the interleukins and nitric oxide synthase to deplete the inflammation response (Sheeba and Asha, 2009). However, the target protein as well as the active compounds with anti-inflammatory profile is still unknown. Rheumatoid arthritis is an inflammatory disease where joints and cartilages progressively diminish and cause pain. Among the different targets that are widely used to reduce chronic inflammation, human DiHydroOrotate Dehydrogenase is widely studied. This target is inhibited by a special class of drugs called Disease Modifying Anti Rheumatic Drug [DMARD]. Currently, many studies are performed to seek out valid alternatives to the drugs available that are practically amenable for long term usage. This precisely is the driving force behind our study. The phytocompounds of *C. halicacabum* were screened, docked and simulated against hDHODH to evaluate the possible inhibitory effects of these compounds on hDHODH.

The 19 compounds in the leaf of *C. halicacabum* were used to screen a validated E pharmacophore specific for an inhibitor of hDHODH. The 5 hits obtained though not of a very high fitness score indicate that desirable features of a good

inhibitor of hDHODH are present in the phytocompounds of *C. halicacabum*. Further the phytocompounds were docked with the receptor of hDHODH to obtain 9 hits having a XP Gscore better than the known inhibitor of hDHODH – A771726. The top hit caftaric acid interacted in a pocket deeper than the known inhibitor site. Caftaric acid although gave very high score of -11.56 Kcal / mol, bound to a FMN binding site rather than the A771726 binding site. This could have serious repercussions as hDHODH is a 2 site catalytic enzyme for both orotate as well as ubiquinone reduction. Thus the binding of caftaric acid to a different site can cross react with oxidative phosphorylation process. The active site residues Arg136 and Tyr356 and the surrounding 3Å amino acids were the main binding site for all the other hits mainly apigenin 7O-glucornide and Luteolin 7O- glucornide, chlorogenic acid and prunigen. The compounds obtained were checked for its credibility by applying additional validation tests like varying docking algorithms, docking with decoys and with other known inhibitors of hDHODH. Further a flexible ligand receptor docking protocol was performed using crossdocking of other known crystal structures of hDHODH to find a lead ligand and its conformational pose.

Thus multiple cross dockings showed majority of the times Luteolin 7O-glucornide as the best hit followed by Apigenin 7O-glucornide. The poses of the cross docked receptor along with the ligand were taken as the starting structure for dynamics. The

simulation time of 10ns was used so as to get an initial preview on the type of interactions and the forces that act against the receptor ligand interactions. Among four complexes simulated, the apo protein had a high RMSD value of 1.59Å compared to the other 3 complexes. The lead compounds Luteolin 7O-glucuronide and Apigenin 7O- glucuronide complexed receptor were comparable and equally good as the known inhibitor A771726 (1.53, 1.52, 1.51 Å resp.) Similar study of this kind has been used to find inhibitors to benign prostatic hyperplasia (Wang *et al.*, 2014) and Low Density Lipoprotein formation (Jiang *et al.*, 2016). Our results indicate that the leaf compounds Luteolin 7O-glucuronide and Apigenin 7O- glucuronide are the principal ingredients of *C. halicacabum* that alleviate that disease of Rheumatoid arthritis by acting on the pyrimidine pathway target of hDHODH. Further, enzyme inhibition studies are warranted.

CONCLUSION

A study to find novel inhibitors for hDHODH was done using E- pharmacophore screening, cross docking and molecular dynamics. All the validations and simulations performed on the complex of hDHODH with the lead-like compounds Luteolin 7O-glucuronide and Apigenin 7O-glucuronide resulted in giving a broad idea on the type of interactions that may occur between *C. halicacabum* leaf compounds and the target hDHODH. This study also provides a foundation for efficiency studies of *C. halicacabum* against a target of Rheumatoid arthritis. The study helps to spell out active ingredients in *C. halicacabum* leaf that may further be used for treatment of Rheumatoid arthritis. Thus Luteolin 7O-glucuronide and Apigenin 7O-glucuronide can be shortlisted as lead compounds which can be optimized for stability. Alternately, our study can be used as a starting point for exploring an alternate mechanism of action of the leaves of *C. halicacabum* against inflammatory targets like hDHODH.

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