Drug Target Identification and Virtual Screening in Pursuit of Phytochemical Intervention of *Mycobacterium chelonae*

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ABSTRACT

Mycobacterium chelonae is a rapidly growing mycobacterium present in the environment. It is associated with skin and soft tissue infections including abscess, cellulitis and osteomyelitis. Other infections by this bacterium are post-operative/transplant-associated, catheter, prostheses and even concomitant to haemodialytic procedures. In this study, we employ a subtractive genomics approach to predict the potential therapeutic candidates, intended for experimental research against this bacterium. A computational workflow was devised and executed to procure core proteome targets essential to the pathogen but with no similarity to the human host. Initially, essential *Mycobacterium chelonae* proteins were predicted through homology searching of core proteome content from 19 different bacteria. Druggable proteins were then identified and N-acetylglucosamine-1-phosphate uridyltransferase (GlmU) was chosen as a case study from identified therapeutic targets, based on its important bifunctional role. Structure modeling was followed by virtual screening of phytochemicals (N > 10,000) against it. 4,4'-[(1E)-5-hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol, apigenin-7-O-beta-gluconopyranoside and methyl rosmarinate were screened as compounds having best potential for binding GlmU. Phytotherapy helps curb the menace of antibiotic resistance so treatment of *Mycobacterium chelonae* infection through this method is recommended.

Key words:

Mycobacterium chelonae, drug design, phytotherapy, GlmU, virtual screening.

INTRODUCTION

Mycobacteria are categorized into two major groups, tubercular and non-tubercular mycobacteria. Nontuberculous mycobacteria are further divided into two groups, rapidly growing and slow growing mycobacteria, depending upon duration of their reproduction in suitable medium (Tortoli, 2014). *Mycobacterium chelonae* is a member of rapidly growing group, which takes less than a week for its reproduction in/on medium. It is the most commonly isolated organism among all rapidly growing mycobacteria. It is mostly found in water sources and on medical instruments such as bronchoscopes (Gonzalez-Santiago and Drage, 2015), and has been isolated from environmental, animal and human sources. *Mycobacterium chelonae* infections in human hosts have increased over time, with reports of both haematogenous and localized occurrences in the recent past (Hay, 2009).

Outbreaks due to *Mycobacterium chelonae* contaminated water and compromised injections are a rising problem. Infections have been linked to cosmetic and surgical procedures, such as trauma, surgery, injection (botulinum toxin, biologics, dermal fillers), liposuction, breast augmentation, under skin flaps, laser resurfacing, skin biopsy, tattoos, acupuncture, body piercing, pedicures, mesotherapy and contaminated foot bath (Gonzalez-Santiago and Drage, 2015). *Mycobacterium chelonae* may also colonize skin wounds, as result of which patients form abscess, skin nodules and sinus tracts (Patnaik *et al.*, 2013).

Recent era has observed a trend for search of drug targets in pathogens using computational methods, with a focus on genomic and proteomic data (Shanmugam and Pan, 2013).

Comparative/differential and subtractive genomics along with proteomics has been used by many researchers for the identification of drug targets in various pathogenic bacteria like *Pseudomonas aeruginosa*, *Helicobacter pylori*, *Brugia malayi*, *Leptospira interrogans*, *Listeria monocytogenes*, *Mycobacterium leprae* (Amineni *et al.*, 2010; Shanmugam and Natarajan, 2010; Sarangi *et al.*, 2015) etc. Determination of potential drug targets has been made possible by the availability of whole genome and their inferred protein complement sequences in public domain databases (Sarangi *et al.*, 2015). Numerous anti-tubercular drug targets, even though mostly for *Mycobacterium tuberculosis* have been reported previously. Some of these include proteins essential for survival or the ones that could be targeted by anti-microbial peptides. Common ones include sulphur metabolic pathway proteins (Reviewed by Bhawe *et al.* 2007) and folate pathway

proteins (Rengarajan *et al.*, 2004) but due to mutations, folate pathway proteins are prone to drug resistance as well.

Parenteral antibiotics against *Mycobacterium chelonae* include tobramycin, amikacin, imipenem, and tigecycline, but it has demonstrated resistance to antibiotics and disinfectants. This property assists *Mycobacterium chelonae* in colonizing water systems and allows its access to humans (Jaén-Luchoro *et al.*, 2016). Successful resolution of *Mycobacterium chelonae* infection with antibiotics has been reported in different scenarios (Brown-Elliott *et al.*, 2001; Choi *et al.*, 2018), but resistance to beta-lactams (Jarlier *et al.*, 1991) and multiple drugs (Churgin *et al.*, 2018) has also been observed. Till now, no specific guidelines for the treatment of *Mycobacterium chelonae* have been defined in the literature (Gonzalez-Santiago and Drage, 2015). Antibiotic resistance also illustrates the need to search out new drug targets for design of better therapies against infection by this bacterium, especially using naturally existing metabolites from microbes and plants. In the current study, subtractive proteomics was applied to identify essential druggable proteins in *Mycobacterium chelonae*. Docking of the selected protein GlmU with phytochemicals was carried out for identification of a candidate which might bind it and stop its normal cellular function, leading to bacterial lysis/death.

MATERIAL AND METHODS

Prediction of *Mycobacterium chelonae* essential proteome

Complete proteome sequence of *Mycobacterium chelonae* CCUG 47445 with accession no: NZ_CP007220 was downloaded from the NCBI database. For the prediction of essential proteins, Geptop (Wei *et al.*, 2013) was installed on computer and a search was carried out to align the *Mycobacterium chelonae* protein sequences against the essential or core protein sequences from defined set of 19 bacteria with an essentiality score cut-off value range of 0.15 (Wei *et al.*, 2013). Results were saved and analyzed further according to the workflow (Fig. 1).

Prediction of non-homologous host proteins

In order to find the bacterial proteins which do not have similarity with human host, the set of essential protein sequences of *Mycobacterium chelonae* was subjected to BLASTP against the human proteome database (Uniprot release 2014). The standalone BLAST software was used for this purpose. For identification of non-homologous proteins, an expectation (E-value) cut-off of 10^{-2} , gap penalty of 11 and gap extension penalty of 1 was set as the standard. E-value cut-off (10^{-2}), based on reported research protocols (Sarangi *et al.*, 2015) was considered.

Identification of putative drug targets

There are several molecular and structural properties which have been explored by researchers for selecting suitable therapeutic targets in pathogenic microorganisms. These properties include determination of molecular weight, sub-cellular localization, 3D structure and druggability analysis. These properties were evaluated for selection of the therapeutic targets in *Mycobacterium chelonae*. Molecular weight was calculated by using computational tools and drug target-associated literature available in the Swiss-Prot database. Subcellular localization of therapeutic targets was predicted using PSORTb (Nancy *et al.*, 2010). It uses feature support vector machine-based method and suffix tree algorithm for downstream analysis. Predictions are grouped through a Bayesian scheme into one final (consensus) result. Druggable targets were identified with BLAST hits through unified protocol from the DrugBank. Parameters were: gap cost: -1 in case of extension or opening, mismatch penalty: -3, E-value: 1×10^{-5} , match: 1, filter algorithm: DUST and SEG (Azam and Shamim, 2014). The targets were subjected to KEGG blast for identification of associated pathways.

Virtual screening of ligand against selected target

Keeping in view the results of sub-cellular localization, molecular weight determination and druggability analysis, an essential protein (GlmU) was chosen for further downstream processing. Swiss Model was used for the prediction of 3D structure of the selected target protein (Biasini *et al.*, 2014). This tool constructs structure model by recognizing structural templates from the PDB using multiple threading alignment approaches (Wang *et al.*, 2016). The top structure used for structure prediction was that of GlmU from *Mycobacterium tuberculosis* (PDB ID: 3D8V). The structure was validated and analyzed for quality using SAVES (<https://services.mbi.ucla.edu/SAVES/>), consisting of ERRAT, VERIFY3D and Ramachandran plot analysis. TM-score was calculated using TM-align (Zhang and Skolnick, 2005).

Phytochemical libraries consisting of three groups were downloaded and docked with GlmU. group I consisted of ~2000 ayurvedic pharmacopia compounds (downloaded from <http://ayurveda.pharmaexpert.ru/>) (Lagunin *et al.*, 2015), group II had ~6000 North African flora compounds (Ntie-Kang *et al.*, 2017) (downloaded from <http://african-compounds.org/nanpdb/>) and group III had 2266 phytochemical compounds from various other medicinal plants (available from <http://bioinform.info/> in .sdf format) (Mumtaz *et al.*, 2016). Blind docking was carried out using Molecular Operating Environment (MOE) with the parameters: placement:

triangle matcher, rescoring 1: London dG, refinement: forcefield, rescoring 2: affinity dG. MOE provides fast and accurate docking results based on dedicated algorithms and accurate scoring functions. Structural preparation program embedded in MOE added the missing hydrogen atoms, corrected the charges and assigned near precise hybridization state to each residue (Basharat *et al.*, 2017). Compounds with least energy and fulfilling the criteria of Lipinski's drug like test were considered for further analysis. ADME/Tox analysis was also performed for the top compounds by inputting the structure at the webserver <https://preadmet.bmdrc.kr/>.

RESULTS AND DISCUSSION

Essential proteome prediction

Initially, total proteome of *Mycobacterium chelonae* was subjected to core or essential proteins prediction. Geptop identified these proteins by screening against 19 bacteria (Supplementary Table 1), based upon orthology and phylogeny features. In the process of essential proteome mining through homology-based methods, a query protein is considered as essential if it is also present in another bacterium and experimentally identified as essential for survival. There are various methods for the prediction of essential proteins, for example single-gene knockout, transposon mutagenesis and RNA interference but all these methods are time consuming and laborious. A good alternative is high-efficiency computational methods designed specifically for this type of work (Wei *et al.*, 2013). Predicted essential proteins were 305 in number (Supplementary Table 2), linked with significant metabolic pathways in the pathogen life cycle and necessary for its survival. In order to disrupt the function and existence of pathogen it is most important to attack those bacterial proteins which regulate important functions (e.g. nutrient uptake) in the host environment (Butt *et al.*, 2012). Latest antimicrobial drugs are designed on the principle of the inhibition of the pathogen's metabolic pathway. Therefore, such protein sequences may be considered as possible therapeutic targets.

Identification of non-host proteins

Non-host proteins refer to those bacterial proteins which do not have homology with human proteins. If the homologous proteins are targeted, they could badly affect the metabolism of host due to similarity with host proteins. Therefore, non-host proteins could be preferred better drug targets, as side effects and cross-reactivity caused by the use of antibiotics could be evaded for harming host (Sarangi *et al.*, 2015). Among the core proteome of *Mycobacterium chelonae*, 117

proteins (Supplementary Table 3) indicated ‘no hit’ against the human proteome according to the set criteria. These proteins were then used for subsequent analysis.

Drug Target mining and analysis

BLASTp was performed to identify significant drug target from newly selected essential proteins. Only 17 proteins had significant hits against druggable proteins present in the DrugBank (Table 1).

Molecular weight determination and druggability analysis could improve the screening process for therapeutic targets, as observed previously for numerous pathogenic bacteria and fungi (Abadio *et al.*, 2011). The molecular weight for each potential drug target was calculated (Table 1) and based upon previous studies, it is suggested that smaller proteins are readily soluble and easier to purify (Duffield *et al.*, 2010).

Sub-cellular localization is a critical factor as it helps in accessing the target gene. Cellular functions are compartment specific, so if the location of unknown protein is predicted then its function could also be known which help in selection of proteins for further study. Membrane proteins are reported as more useful target and more than 60% of the currently known drug targets are membrane proteins (Tsirigos *et al.*, 2015). Cytoplasm is the site of proteins synthesis and most of these proteins remain there to carry out their specific functions after synthesis. However, some proteins need to be transported to different cellular compartments for their specific function (Strzyz, 2016). The subcellular localization of non-host proteins of *Mycobacterium chelonae* was predicted and majority were demarcated as cytoplasmic (Table 1).

GlmU analysis and phytochemical screening

After the characterization of all druggable proteins of *Mycobacterium chelonae*, GlmU was selected for further analysis (out of 305 essential and 117 non-homologous proteins). It is a bifunctional enzyme, exhibiting both acetyltransferase and uridyltransferase activities (Sharma *et al.*, 2016) and involved in biosynthesis of peptidoglycan. Peptidoglycan in complex with mycolyl-arabinogalactan is important for the viability of Mycobacteria as it upkeeps the basal structure necessary for upper “myco-membrane” support. GlmU has been analyzed for various bacterial species, including *Escherichia coli*, *Streptococcus pneumonia*, *Haemophilus influenzae* and *Mycobacterium tuberculosis* (Patin *et al.*, 2015). Predicted structure of GlmU had an estimated RMSD of 0.5 Å from the reference structure. A TM-score of 0.97 was obtained and since a score of 0.5-1 means same fold, with greater value meaning more similarity. 0.97 meant a

very similar fold structure. Structure was found to be a homotrimer. Ramachandran plot showed 97.2% residues in favored regions (90.4% in core, 9.4% in allowed, 0.3% in generously allowed) and no residue in disallowed region (Fig. 2). According to VERIFY3D program, at least 80% of the amino acids should have value ≥ 0.2 in the 3D/1D profile and 96.55% of predicted GlmU residues had an averaged 3D-1D score ≥ 0.2 , thus passing the quality check. An ERRAT score of 90.66 was obtained.

Each monomer of GlmU consists of two domains: N and C-terminal domains. N-terminal domain has α/β like fold, similar to dinucleotide-binding Rossmann fold topology. C-terminal domain exhibits a regular left-handed β -helix conformation and a long α -helical arm connecting both domains (Sharma *et al.*, 2016). N-terminal domain is essential for uridyltransferase activity as it catalyses the transfer of uridine monophosphate from uridine-triphosphate to N-acetylglucosamine-1-phosphate (GlcNAc1P). C-terminal domain has acetyltransferase activity as it catalyzes the transfer of an acetyl group from acetyl-CoA coenzyme to GlcN1P, in order to produce GlcNAc1P (Moraes *et al.*, 2015).

GlmU plays fundamental role in the formation of bacterial cell wall by carrying out catalysis of uridine-diphospho-N-acetylglucosamine, an important precursor in bacterial peptidoglycan cell wall (Sharma *et al.*, 2016). Activity of acetyltransferase of eukaryotic cells differ from the activity of GlmU in the way that, eukaryotic acetyltransferase occurs on GlcN6P and not on GlcN1P. These properties make GlmU suitable as the drug target (Moraes *et al.*, 2015) and it has been used for drug targeting in bacteria such as *Haemophilus influenza* (Mochalkin *et al.*, 2007) apart from designing inhibitors against it for Mycobacterial species (Mehra *et al.*, 2016). It is also known that proteins that are involved in more than one pathway of pathogen, in addition to that they are non-host proteins, could be more effective drug targets (Sarangi *et al.*, 2015). Inactivation of bifunctional GlmU enzyme leads to loss of mycobacterial viability, therefore GlmU was used for docking against phytochemicals.

The possible interaction between protein and the ligand is understood computationally through molecular docking therefore receptor centric docking approach was employed for screening of prospective phytochemical library of compounds from more than 10,000 plant derived compounds against GlmU of *Mycobacterium chelonae*. Whole surface of the protein was scanned for best docking pose. A comparative analysis of structural shape and chemical complementarity to GlmU was ranked, based on S value and Lipinski's drug likeliness in MOE.

Ones with least S-value and Lipinski's drug likeliness=1 were considered as best inhibitors. Results of GlmU binding with compounds from group I, II and II showed several interactions (Fig. 3, Table 2). 4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol from group I, apigenin-7-O-beta-gluconopyranoside (also commonly known as apigetrin) from group II and methyl rosmarinate was obtained as the best anti-microbial against *Mycobacterium chelonae* from group III compounds. Common source of 4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol is *Alpinia galanga* while apigetrin is sourced from dandelion coffee and *Teucrium gnaphalodes*. Methyl rosmarinate is a constituent of *Rabdosia serra*, *Ehretia thyrsoflora*, *Dracocephalum heterophyllum*, *Perilla frutescens*, *Nepeta deflersiana*, *Cordia sinensis* and *Hyptis atrorubens* Poit. ADME and toxicity parameters were also calculated and it was found that apigenin-7-O-beta-gluconopyranoside had lowest blood brain penetration but predicted as a mutagen while methyl rosmarinate and 4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol were found to be non-mutagenic and had slightly higher blood brain barrier penetration (Table 3). Analogs of apigenin have shown activity against *Mycobacterial* species (Tsouh Fokou *et al.*, 2015; Evina *et al.*, 2017) while other compounds have not been reported in literature as anti-mycobacterials yet.

We focused on phytochemical screening against *Mycobacterium chelonae* GlmU because plant derived/natural compounds could be used as antibacterial therapeutics for treatment of bacterial infections. Further Lab testing is proposed to know about minimum inhibitory concentration value and other parameters for these compounds against *Mycobacterium chelonae*.

CONCLUSION

In this study, we proposed GlmU as one of the important therapeutic drug targets for *Mycobacterium chelonae* as it is bifunctional, essential protein for pathogen and has no homology with human proteome. We have provided putative model for phytotherapy against *Mycobacterium chelonae* through virtual screening-based identification of potent metabolites from three data sets, comprising of more than 10,000 compounds. This study could be taken as an initiative for screening and quick designing of phytotherapy against microbes using a computational *modus operandi*. It is expected that our study will also facilitate selection and screening of other *Mycobacterium chelonae* therapeutic target proteins against phytochemical and other relevant compounds for Lab testing and successful entry into drug design pipeline.

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DECLARATION OF INTEREST

The authors declare that they have no conflict of interest.

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FIGURE/TABLE LEGENDS

Fig. 1. Workflow describing subtractive proteomics with respective number of obtained sequences. *In silico* parameter analysis included hydrophobicity, sub-cellular location prediction, molecular weight etc.

Fig. 2. Ramachandran plot of the predicted GlmU structure.

Fig. 3. (A) Predicted 3D structure of GlmU **(B)** Docked GlmU with 4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol showing residue interactions (2D conformation) **(C)** apigenin-7-O-beta-guconopyranoside and GlmU docked complex(2D conformation) **(D)** Methyl rosmarinate and GlmU docked complex(2D conformation). Blue arrowhead represents backbone donor/acceptor while green arrowhead depicts sidechain donor/acceptor, dotted line represents proximity contour.

Table 1. Predicted druggable proteome of *Mycobacterium chelonae*. GRAVY is average hydrophobicity of the protein measured by Kyte-Doolittle algorithm. Hydrophobicity value below 0 are indicates globularity (means protein is hydrophilic), while score value of above 0 indicates protein to be membranous (hydrophobic).

Table 2. Top 5 compounds against GlmU of *Mycobacterium chelonae* (from each docked group obtained with best inhibition score).

Table 3. ADME and toxicity profile of top drug-like molecules from studied phytochemical group compounds.

Supplementary Table 1. List of bacterial species used for essential proteome prediction of *Mycobacterium chelonae*.

Supplementary Table 2. Essential proteome of *Mycobacterium chelonae*.

Supplementary Table 3. Non-homologous core proteome of *Mycobacterium chelonae*.

Fig. 1.

Fig. 2.

Fig. 3.

Table 1. Predicted druggable proteome of *Mycobacterium chelonae*. GRAVY is average hydrophobicity of the protein measured by Kyte-Doolittle algorithm. Hydrophobicity value below 0 are indicates globularity (means protein is hydrophilic), while score value of above 0 indicates protein to be membranous (hydrophobic).

S. No.	Accession #	Protein name	No. of Matches	Mol. Wt.	Length (amino acids)	GRAVY	Sub cellular localization	KEGG Pathway	Experimental evidence for druggability in bacteria
	WP_030093515	6,7-dimethyl-8-ribityllumazine synthase/riboflavin synthase	1	16775.9	162	0.173	Unknown	Riboflavin metabolism, biosynthesis of secondary metabolites	Zhao <i>et al.</i> , 2009
	WP_030095371	acyl carrier protein	1	10805.0	99	-0.207	Cytoplasmic	Fatty acid biosynthesis	Luckner <i>et al.</i> , 2010
	WP_030097637	replicative DNA helicase	1	50372.3	458	-0.323	Cytoplasmic membrane	DNA replication	Aiello <i>et al.</i> , 2009
	WP_046251970	DNA topoisomerase IV subunit B	3	74147.8	675	-0.371	Cytoplasmic	DNA repair and recombination	Li <i>et al.</i> , 2016
	WP_046252393	dihydropteroate synthase	2	37365.3	350	-0.165	Cytoplasmic	Folate biosynthesis	Zhao <i>et al.</i> , 2016
	WP_046252395	2-amino-4-hydroxy-6-hydroxymethyl dihydropteridine diphosphokinase	1	19332.9	178	-0.054	Cytoplasmic	Tetrahydrofolate biosynthesis	Shi <i>et al.</i> , 2001
	WP_046252399	pantoate - beta-alanine ligase/ pantothenate synthetase	1	33527.3	314	-0.057	Cytoplasmic	Pantothenate biosynthesis	Yang <i>et al.</i> , 2011
	WP_046252764	3-oxoacyl-ACP synthase/beta-ketoacyl acyl carrier protein synthase	2	35485.1	340	0.070	Cytoplasmic	Lipid synthesis	Lee <i>et al.</i> , 2012
	WP_046252769	bifunctional N-acetylglucosamine-1-phosphate uridylyltransferase (GlmU)/glucosamine	1	50208.3	482	-0.026	Cytoplasmic	Amino sugar and nucleotide sugar metabolism, Biosynthesis	Mehra <i>et al.</i> , 2015

		ne-1-phosphate acetyltransferase						of antibiotics	
	<u>WP_0462</u> <u>53413</u>	cell division protein FtsI	1	68842.9	642	-0.300	Cytoplas mic membrane	Peptidoglycan biosynthesis, beta-Lactam resistance	Slayden and Belisle, 2009
	<u>WP_0462</u> <u>53421</u>	cell division protein FtsZ	1	38795.6	383	0.076	Cytoplas mic	Chromosome partition, cytoskeleton, toxin-anti- toxin	Park <i>et al.</i> , 2014
	<u>WP_0462</u> <u>53835</u>	enoyl-[acyl- carrier-protein] reductase (InhA)	3	28724.0	269	0.152	Cytoplas mic membrane	Lipid biosynthesis	Luckner <i>et al.</i> , 2010
	<u>WP_0462</u> <u>54273</u>	DNA ligase (NAD(+)) LigA	1	74430.9	684	-0.226	Cytoplas mic	DNA replication, Base excision repair, nucleotide excision repair, mismatch repair	Mills <i>et al.</i> , 2011
	<u>WP_0462</u> <u>54500</u>	thymidylate kinase	1	23403.1	212	-0.394	Cytoplas mic	Pyrimidine metabolism	Kosinska <i>et al.</i> , 2005
	<u>WP_0462</u> <u>54518</u>	biotin--[acetyl- CoA-carboxylase] ligase	1	27934.5	264	-0.084	Cytoplas mic	Biotin metabolism	Soares da Costa <i>et al.</i> , 2012
	<u>WP_0462</u> <u>54614</u>	alanine racemase	1	41861.3	397	-0.008	Cytoplas mic	D-Alanine metabolism, Vancomycin resistance	Anthony <i>et al.</i> , 2011
	<u>WP_0462</u> <u>55923</u>	ribonucleotide- diphosphate reductase subunit beta/ ribonucleoside diphosphate reductase-β	2	36899.2	320	-0.437	Unknown	Purine and pyrimidine metabolism	Tholander and Sjöberg, 2012

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348
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Table 2. Top 5 compounds against GlmU of *Mycobacterium chelonae* (from each docked group obtained with best inhibition score).

Group I		Group II		Group III	
Compound	S-value	Compound	S-value	Compound	S-value
4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol	-6.63	Apigenin-7-O-beta-glucuronopyranoside	-6.35	Methyl rosmarinate	-12.33
1-(3,4-dihydroxy-5-prenylphenyl)-2-(3,5-dihydroxyphenyl)ethane	-6.49	Cosmetin	-5.88	5,7,3',4',5'-Pentahydroxyflavanone	-12.23
4,4',7,9-tetrahydroxy-8->9'-neolign-7'-ene	-6.35	Anastatin B	-5.77	Peonidin	-11.95
1,7-bis(3,4-dihydroxyphenyl)-1,4,6-heptatrien-3-one	-5.79	Ixerisoside D	-5.66	Flowerone	-11.92
Nimocinolide	-5.77	Quercetin	-5.59	Hovenitin III	-11.82

Table 3. ADME and toxicity profile of top drug-like molecules from studied phytochemical group compounds.

Parameter	4,4'-[(1E)-5-Hydroxy-4-(methoxymethyl)-1-pentene-1,5-diyl]diphenol	Apigenin-7-O-beta-gluconopyranoside	Methyl rosmarinate
<i>In vivo</i> blood-brain barrier penetration	1.90	0.03	0.17
CYP450 2D6 inhibition	None	None	None
Human intestinal absorption (%)	90.47	47.10	76.88
Plasma protein binding (%)	89.08	73.43	86.57
Ames test	Non-mutagen	mutagen	Non-mutagen
Skin permeability (logKp, cm/hour)	-2.79	-4.52762	-3.30
hERG inhibition	Medium risk	High risk	Medium risk
Carcinogenicity in mouse	Negative	Positive	Negative
Carcinogenicity in rat	Negative	Negative	positive

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