NATURAL PRODUCTS: FROM CHEMISTRY TO PHARMACOLOGY (C HO, SECTION EDITOR)



Ellagic Acid, Kaempferol, and Quercetin from *Acacia nilotica*: Promising Combined Drug With Multiple Mechanisms of Action

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Abstract

The pharmacological activity of *Acacia nilotica*'s phytochemical constituents was confirmed with evidence-based studies, but the determination of exact targets that they bind and the mechanism of action were not done; consequently, we aim to identify the exact targets that are responsible for the pharmacological activity via the computational methods. Furthermore, we aim to predict the pharmacokinetics (ADME) properties and the safety profile in order to identify the best drug candidates. To achieve those goals, various computational methods were used including the ligand-based virtual screening and molecular docking. Moreover, pkCSM and SwissADME web servers were used for the prediction of pharmacokinetics and safety. The total number of the investigated compounds and targets was 25 and 61, respectively. According to the results, the pharmacological activity was attributed to the interaction with essential targets. Ellagic acid, Kaempferol, and Quercetin were the best *A. nilotica*'s phytochemical constituents that contribute to the therapeutic activities, were non-toxic as well as non-carcinogen. The administration of Ellagic acid, Kaempferol, and Quercetin as combined drug via the novel drug delivery systems will be a valuable therapeutic choice for the treatment of recent diseases attacking the public health including cancer, multidrug-resistant bacterial infections, diabetes mellitus, and chronic inflammatory systemic disease.

 $\textbf{Keywords} \ \textit{A. nilotica} \cdot \text{Ellagic acid} \cdot \text{Kaempferol} \cdot \text{Quercetin} \cdot \text{Multiple mechanisms of action} \cdot \text{ADMET} \text{and computer-aided drug discovery}$

Introduction

Acacia nilotica is a tropical and sub-tropical medicinal plant belonging to the Fabaceae family [1]. No doubt, medicinal plants play a vital role in drug discovery, since they are affluent with bioactive phytochemical constituents that are valuable in the treatment of various diseases, particularly those causing recent threats attacking the public health including cancer, multidrug-resistant bacterial infections, diabetes mellitus, and chronic inflammatory systemic diseases [2, 3].

The higher incidence of cancer and mortality rate [4], the emergence of bacterial resistance with the declining in the

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antibacterial research at several pharmaceutical companies [5], the huge prevalence and complications associated with diabetes mellitus [6] as well as the long-term suffering associated with the chronic inflammatory systemic diseases such as rheumatoid arthritis and multiple sclerosis [7] are leading forces that encourage us to participate in fighting against the probable threats. Such an issue is attained via the discovery and development of efficient innovative anticancer, antibacterial, antidiabetic, and anti-inflammatory drugs. Unfortunately, drug discovery is a time-consuming, costive, as well as difficult process [8, 9]; hence, it necessitated to involve sophisticated techniques in the drug discovery process in order to overcome those limitations. Recently, one of the promising sophisticated techniques is the computational tools (computer-aided drug design) that have a valuable impact in the discovery and development of newer drugs with a reduction in time and cost [8]. They include the ligand-based virtual screening that is based on the searching for the compounds having the highest probability in pharmacological activity [10] and molecular docking that relies on the energy-based scoring function to identify ligand-target complex lowest



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energy [11]. Moreover, they involve the software of pharmacokinetics, toxicity, and the drug-likeness prediction that work by many algorithms [12] including the graph-based signature [13]. Many studies concerning the application of the computational tools in the discovery of natural-derived drugs were conducted [14–17].

A. nilotica is opulent of many phytochemical constituents including tannins, alkaloids, terpenoids, and flavonoids. Many studies were conducted in it resulting in an evidence-based pharmacological data that revealed the potential pharmacological activities of the phytochemical compounds including anticancer, antibacterial, antidiabetic, anti-inflammatory, and other activity making the plant as a promising source for the development of innovative, safe, biodegradable drugs with great activity. The chemical structure of active A. nilotica's phytochemical constituents was elucidated, the correlation between the responsible phytochemical constituents for treatment and the diseases were conducted [1], but the determination of exact targets that phytochemical constituents bind and the mechanism of action were not performed; consequently, based on established literature and studies, we aim to identify the exact targets that phytochemical constituents bind to exert the pharmacological activity by utilizing the computational methods as a tool for the study so as to understand the mechanism of action. Within the current drug design pipeline, drug target identification is a very important step in the understanding of the probable mechanism of action, increasing the confidence and reducing the attrition in clinical trials [18]. Furthermore, we aim to predict the pharmacokinetics (ADME) properties and safety profile with the intention of identifying the best drug candidates. The QSAR-based virtual screening is characterized by great and fast throughput with respectable hit rank [10]. Molecular docking is valuable to predict the stability of the ligand-target complex that reflects the biological activity [19]. The pharmacokinetics, toxicity, and drug-likeness prediction are helpful to identify the best drug candidates [12, 20]. To our knowledge, such a study was not conducted before.

Materials and Methods

Ligand-Based Virtual Screening

Ligands Preparation

The chemical structure of the reported *A. nilotica*'s phytochemical constituents (25 compounds) [1] was drawn via Marven Sketch software version 18.5 [21] (Fig. 1). The 3D structure was generated in a mol2 format with Open Babel software [22], minimized and optimized with Cresset Flare software [23] at the accurate type calculation method.



Virtual Screening

The screening for the exact target that the phytochemical constituents bind was performed via Similarity Ensemble Search Tool [24] and TargetNet web servers [25]. The compound structures were submitted in smile format. The targets with higher probability score were selected for further validation via molecular docking study (61 targets). The linkage between predicted targets with the diseases was attained via UniProt [26], Pharos [27], and Therapeutic Target Databases [28]. The results are listed in Tables 1, 2, 3, 4, 5, and 6.

Molecular Docking

Target Preparation

The 3D structure of selected targets from virtual screening was obtained from the RCSB protein data bank [67]. The structure with better resolution and validation scores was selected for the study. In order to validate the docking results, multiple 3D X-ray crystallographic structures for the same target were downloaded in PDB format. For the structures that have no practically determined 3D structure, Phyre2 [58], SWISS-MODEL web server [49], and RaptorX [59] web servers were used for 3D structure modeling, then downloaded in PDB format. The target preparation was carried out in Cresset Flare software [23] according to the default settings. After preparation, the targets 3D structures were minimized via Cresset Flare software [23] at the normal type calculation method. The targets were input to the software in PDB format.

Ligands Preparation

The preparation of reported *A. nilotica*'s phytochemical constituents for molecular docking study was carried out as described above.

Molecular Docking of Phytochemical Constituents With the Predicted Targets

The docking calculations were carried out in Cresset Flare software [23] in normal mode and default settings. The grid box was defined according to the co-crystallized ligands, but in the absence of co-crystallized ligands, the grid box was defined via picking of active site amino acids. Beside the *A. nilotica*'s phytochemical constituents, drugs that are well known to bind with the predicted targets (selected randomly from Therapeutic Target [28] and Pharos [27] databases) and the co-crystallized ligands were used as positive controls. The compounds and the targets were input in mol2 and PDB format, respectively.

The results are listed in Tables 1, 2, 3, 4, 5, 6, and 7 and Figs. 1, 2, 3, 4, and 5.

The Pharmacokinetics and Toxicity Prediction

The intestinal absorption, volume of distribution, bloodbrain barrier, p-glycoprotein and cytochrome-P enzymes inhibition, the renal OCT2 substrate probability, and total clearance were predicted via pkCSM [13] and SwissADME web servers [12]. Moreover, the hepatotoxicity, skin sensitization, the hERG potassium channel inhibition, AMES toxicity, human maximum tolerated dose, carcinogenicity, oral rate acute, and chronic toxicity were predicted via pkCSM web server [13] at the default settings via submitting of the chemical structures in smile format. The results of pharmacokinetics are listed in Tables 7 and 8 and toxicity in Table 9.

Drug-Likeness and Medicinal Chemistry Friendliness Prediction

The probability of *A. nilotica*'s phytochemical constituents to be as drug candidates was carried via applying of Lipinski, Ghose, Veber, Egan, and Muegge filters. In addition, lead likeness and synthetic accessibility were used to predict medicinal chemistry friendliness. The prediction was carried out via SwissADME web server via submitting of the chemical structures in smile format [12]. The results are listed in Table 10.

Methods Validation

The consistency and the reproducibility of the used tools including the molecular docking were validated by the resubmission of the compounds for many times.

Results

The Anticancer Targets

The anticancer activity of *A. nilotica*'s was attributed to the suppression of the oncogenic transformations, progression, and development, DNA replication, and transcription. Moreover, the prevention of cancer cells proliferation, invasion, angiogenesis as well as the suppression of drug resistance and the induction of apoptosis.

The anti-breast cancer activity was due to the inhibition of the aromatase enzyme and estrogen receptor beta. In contrast, the anti-prostate cancer activity is due to the control of metastatic behavior of prostate cancer via the interaction with nuclear receptor ROR-alpha and the

inhibition of Steroid 17 alpha-hydroxylase (Table 1 and Fig. 2).

The Antibacterial Targets

The antibacterial activity of *A. nilotica* was attributed to the prevention of fatty acids, peptidoglycans biosynthesis as well as the prevention of bacterial resistance to the beta-lactam antibiotics. The fatty acid biosynthesis inhibitory activity was against different types of bacteria including *Mycobacterium*, *Pseudomonas aeruginosa*, and *Vibrio cholera*.

The Antiviral and Antiplasmodial Targets

The antiviral activity was attributed to the action on toll-like receptor 9. The anti-HIV activity is due to the inhibition of HIV integrase enzyme. The anti-coronavirus activity is due to coronavirus replicase polyprotein 1 ab enzyme. The antiplasmodial activity was attributed to the inhibition of enzymes MO15-related protein kinase pfmrk and M18 aspartyl aminopeptidase as well as the prevention of fatty acid biosynthesis via inhibition of the enzymes: β -hydroxy acyl-ACP dehydratase FabZ and hydroxyacyl-[acyl-carrier-Protein] dehydratase (Table 3).

The Antidiabetic Targets

The antidiabetic activity was attributed to the interaction with the insulin receptor, glycogen phosphorylase enzyme, sodium/glucose co-transporter 2 as well as the aldose reductase enzyme (Table 4).

The Anti-Inflammatory Targets

The anti-inflammatory activity was attributed to the inhibition of enzymes: arachidonate 15-lipoxygenase, cyclooxygenase-2 (COX-2), phospholipase A2, receptor-interacting serine/threonine protein kinase 2, and xanthine dehydrogenase/oxidase as well as the interaction with macrophage migration inhibitory factor (Table 5).

The Antidiarrheal, Anti-Platelets, and Anticholinesterase Targets

The antidiarrheal activity was attributed to the interaction with the opioid receptors Mu-type Delta-type. The Anti-platelets activity was attributed to the interaction with the P2Y12 receptor. The inhibition of the enzyme acetylcholinesterase is contributed to the anticholinesterase activity.



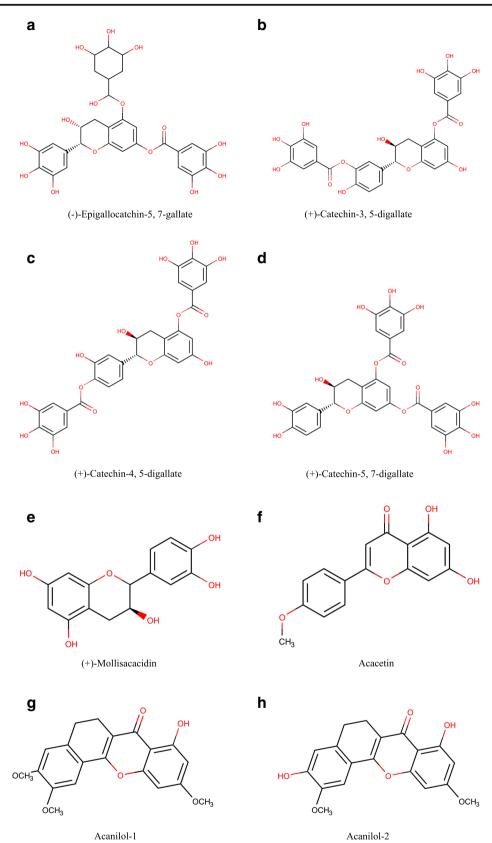


Fig. 1 a–x The 2D chemical structures of the reported A. nilotica's phytochemical constituents [1]



Fig. 1 (continued)

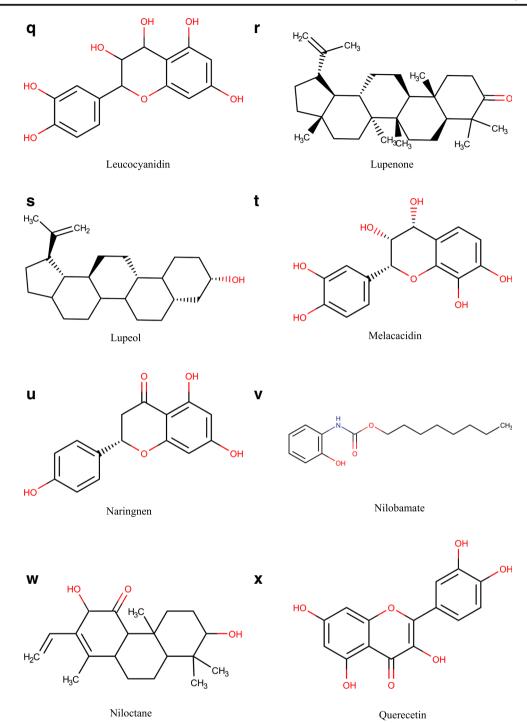


Fig. 1 (continued)

The Predicted Pharmacokinetic Properties

According to the results, Acacetin, γ -Sitosterol, Kaempferol, Flavone, Lupenone, Lupeol, Niloctane, and Quercetin had the highest gastrointestinal absorption, tissue distribution (Vd), and respectable total clearance.

Moreover, Flavone, Nilobamate, and Niloctane were permeable to the blood-brain barrier (BBB). Besides, acanilol-1,

acanilol-2, γ -sitosterol, flavone, lupenone, and lupeol were found to be subjected to the metabolism via CYP3A4 enzyme (Table 7).

Moreover, (+)-Mollisacacidin, Catechin, Chalconaringnen-4-O-beta-glucopyranoside, Epicatechin, Niloticane, Kaempferol-7-glucoside, Leucocyanidin, and Nilobamate were free from drugdrug interaction via the inhibition of cytochrome-P (CYP) or P-glycoprotein (P-gp) I and II enzymes (Table 8).



 Table 1
 Ligand-based virtual screening and molecular docking results regarding the anticancer activity

Target	Compounds	LBVS	Docking score		
1- Anaplastic lymphoma kinase enzyme			4FNZ*	2XB7*	
"It is pivotal in neural cells proliferation	1- Quercetin	1.0	-8.293	-8.31	
and survival" [26]	a- Staurosporine	_	-9.306	10.95	
	• A NZF 1501	_	-7.729	_	
	• A GUI 1501	_	_	- 10.6	
2- Angiopoietin 1 receptor			3L8P*	4X3J*	
"It regulates the angiogenesis, adhesion,	1- Ellagic acid	1.0	-8.403	-7.48	
cell spreading and the maintenance of	a- Cabozantinib	-	-9.041	-7.36	
vascular quiescence" [28]	• A 0CE 2207	_	-13.37	_	
	• A 3WR 1201	-	_	-8.56	
3- Aromatase enzyme			3EQM*	5JKW*	
"It is involved in the estrogen	1- (+)-Mollisacacidin	1.0	-7.840	-8.48	
biosynthesis via conversion	2- Acacetin	0.74	-8.317	-8.44	
of androgens into estrogens" [26]	3- Catechin	1.0	-8.795	-8.41	
	4- Ellagic acid	1.0	-8.664	-8.65	
	5- Epicatechin	1.0	-8.028	-8.45	
	6- Flavone	1.0	-6.450	-6.45	
	7- γ -Sitosterol	1.0	- 11.51	-11.4	
	8- Kaempferol	1.0	-7.406	-7.78	
	9- Leucocyanidin	0.99	-11.77	-12.1	
	10- Naringenin	1.0	-8.354	-8.04	
	11- Quercetin	1.0	-7.579	-8.42	
	a- Anastrozole	_	-7.142	-8.54	
	• A ASD 601	_	-9.699	_	
	• A TES 601	_	_	-9.889	
4- ATP binding cassette sub-family			5NJ3*	6FEQ*	
G member	1- (+)-Catechin-3,5,-gallate	0.95	- 10.01	-11.17	
"Has a major role in cancer's cells	2- Acacetin	1.0	- 5.343	-7.633	
multi-drug resistance" [26]	3- Chalconaringnen-4-O-	0.98	-7.067	-10.81	
mana arag resistance [20]	beta glucopyranoside	0.50	7.007	10.01	
	4- Ellagic acid	1.0	- 5.789	-8.654	
	5- Kaempferol	1.0	-5.839	-8.052	
	6- Leucocyanidin	1.0	- 7.817	-10.37	
	7- Naringenin	1.0	-6.023	-7.732	
	8- Querecitin	1.0	- 5.979	-8.520	
	• A NAG 702	1.0	- 5.220	- 8.320 -	
			- 3.220 -	- - 14.81	
5 A A 1 D 1.:	• A D6T 1002	_			
5- Aurora A and B kinase enzymes	1 FII	1.0	5ORR*	4B8L*	
"Aurora A is involved in regulation	1- Ellagic acid	1.0	-8.327	-8.087	
of cell cycle progression"	2- Quercetin	1.0	- 6.627	-8.572	
"Aurora B is a main regulator	a- Axitinib	_	- 0.400	-6.822	
of mitosis" [16]	• A ADP 401	_	- 9.409	- 202	
C D 10 1 1 1 1 1 1	• A A0P 1352	_	_	-9.203	
6- Bcl-2-related protein A1	1 () 3 (11)	0.04	5WHH*	2VM6*	
"Pro and anti-apoptotic protein" [29]	1- (+)-Mollisacacidin	0.94	- 7.448	-7.925	
	2- Catechin	0.95	- 7.922	- 8.905	
	3- Catechinn-7-O-gallate	0.95	- 7.888	-8.374	
	4- Ellagic acid	1.0	- 8.046	-7.130	
	5- Epicatechin	0.93	-7.340	-8.150	
	6- Kaempferol	1.0	-7.493	-7.061	
	7- Leucocyanidin	0.976	-8.616	-9.055	
	8- Naringenin	0.981	- 7.774	-7.350	
	9- Querecitin	1.0	- 8.516	-8.115	
	a- Venetoclax	-	-9.339	-8.807	
7- Caspase 9 enzyme			1JXQ*	2AR9*	
"Is an apoptotic initiator, acting as	1- (-)-Epigallocatchin-7-gallate	1.0	-11.834	-13.18	
an important therapeutic target" [30]	2- Ellagic acid	0.998	-8.206	-9.408	
	3- Flavone	0.90	-6.693	-7.173	
	4- Gallic acid	1.0	-6.848	-7.562	
	5- Kaempferol	0.846	-8.078	-8.158	
	6- Quercetin	1.0	-9.214	-8.518	
	a- Isatin sulfonamide 34	_	- 10.425	-9.046	
8- Cell division control protein 42			5UPK*	4YDH*	
homology (CDC42)	1- Acacetin	0.999	-8.152	-8.270	
GJ \ /	2- Acanilol-1	-	-8.348	-8.431	



Table 1 (continued)

Target	Compounds	LBVS	Docking score	•
"It participates in the oncogenic	3- Acanilol-2	_	-8.659	- 8.754
transformation,	• C GNP 200	_	- 13.968	_
invasion, and tumorigenesis [31]	• B GNP 201	_	_	-14.66
9- Cell division cycle-7-related protein kinase			4F9B*	4F99*
'It is required to initiate the DNA replication" [32]	1- Ellagic acid	0.993	-8.517	-8.564
	2- Flavone	0.90	-6.085	-6.069
	• C 0SY 601	-	- 12.593	_
	• A ADP 601	-	-	-8.397
10- Cyclin-dependent kinase 1 enzyme			4YC6*	Swiss
"It is a critical regulator of cell cycle progression" [33]	1- Ellagic acid	0.993	− 7.957	-8.845
11- Cyclin-dependent kinase CDK 4 and 6			2W9Z*	4AUA*
'CDK4 regulates the cell cycle during G (1)/S	1- Acacetin	1.0	-7.428	-8.247
transition" CDK 6 promotes G	2- Acanilol-1	_	-7.491	-9.171
(1)/S transition [18]	3- Acanilol-2	_	-7.413	-9.342
	a- Fascaplysin	_	-6.764	-9.109
	• A 4AU 1302	_	_	-6.659
12- Cyclin-dependent kinase 9 enzyme			4BCG*	6GZH*
'It is involved in regulation of	1- Acacetin	1.0	-7.600	-7.785
Transcription" [18]	2- Acanilol-1	_	-8.064	-8.435
	3- Acanilol-2	_	-8.852	-9.146
	• A T3C 1328	_	-8.724	_
	• A LCI 2001	_	-	-9.039
13- Death-associated protein kinase 1	71 ECT 2001		5AUV*	5AUU*
'It regulates type I Apoptotic, type II	1- Quercetin	1.0	-8.275	-10.389
autophagic cell deaths" [16]	2- Kaempferol	1.0	-8.106	- 8.955
autophagic cen deaths [10]	• A AGI 400	-	- 7.349	- 6.733
	• A LU2 400	_	- -	-10.16
14- DNA topoisomerase 1 enzyme	A LO2 400		1SC7*	1TL8*
	1 16 di O collevil hata	0.987		-8.781
"Has an important function in	1- 1,6-di-O-galloyl-beta2- Acacetin	0.987	- 14.08	-6.761
DNA replication" [34]		1.0	- 10.42	-9.436
	3- Kaempferol-7-gluc		- 14.05	- 7.811
	a- Camptothecin	_	- 12.86	
	b- Edotecarin	_	- 16.28	- 17.452
	• C M38 990	_	- 10.74	7,000
16 E 1 ' 4 B 4 4	• D AI3 901	_	- 277774*	-7.088
15- Ephrin type B receptor 4	4 799 1 11	4.0	3ZEW*	6FNI*
'It is important in tumor angiogenesis" [16]	1- Ellagic acid	1.0	- 9.448	-9.537
	• A STU 1889	_	- 11.729	-
46.5	• A DXH 1001	_	-	- 13.698
16- Estrogen receptor beta			1X7J*	2NV7*
'has a pivotal role in the development	1- (+)-Mollisacacidin	1.0	- 9.685	-9.507
and the progression of tumors	2- Acacetin	1.0	-7.556	-7.642
via the mitogenic	3- Catechin	1.0	- 9.966	-9.551
action of estrogens" [35]	4- Catechin-7-O-gallate	1.0	-9.926	-9.978
	5- Dicatechin	1.0	31.247	41.893
	6- Epicatechin	1.0	-9.975	-9.723
	7- Ellagic acid	1.0	-9.212	-8.646
	8- γ-Sitosterol	1.0	- 1.256	-2.088
	9- Kaempferol	1.0	-8.469	-9.027
	10- Leucocyanidin	1.0	-8.452	-4.881
	11- Naringenin	1.0	-9.140	-9.363
	12- Querecitin	1.0	-8.862	-9.456
	• A GEN 201	_	-9.001	_
	• A 555 501	_	_	-9.330
17- Focal adhesion kinase enzyme			4K9Y*	4D58*
'It is essential in angiogenesis, cell	1- Ellagic acid	1.0	-8.510	-9.052
migration and apoptosis" [16]	2- Quercetin	1.0	-9.228	-7.587
- 11 - 1	• A K9Y 701	_	- 9.549	_
	• B BI9 1690	_	_	-8.303
18- Glycogen synthase kinase 3 β	222 10,0		4IQ6	5K5N
'It phosphorylates various proteins i	1- Ellagic acid	1.0	- 9.049	-8.588
n the cell cycle and apoptosis.	• B 6QH 401	-	9.0 4 9	-6.632
Its inhibitors promote apoptosis" [36]	• B IQ6 501		- - 8.067	- 0.032
19- Inducible nitric oxide synthase	- D 100 201	_		3NQS*
13- moncinie mine oxide symmase			3E7G*	3NUS*



Table 1 (continued)

Target	Compounds	LBVS	Docking score		
"It produces NO that has a tumoricidal	1- (+)-Mollisacacidin	1.0	- 10.097	- 9.977	
action in macrophage" [16]	2- Epicatechin	1.0	-10.205	-9.723	
1 2 1 3	3- Ellagic acid	1.0	-8.337	-8.101	
	4- Kaempferol	0.968	-9.407	- 9.549	
	5- Lupenone	1.0	-10.186	-10.53°	
	6- Lupeol	1.0	-12.137	-10.354	
	7- Naringenin	0.89	-9.896	-9.918	
	8- Niloticane	1.0	-10.340	-10.484	
	9- γ-Sitosterol	1.0	-11.662	- 10.763	
	10- Quercetin	1.0	-9.456	- 9.771	
	a- Curcumin	-	-10.024	-9.775	
	• A AT2 906	_	-8.470	-8.788	
20- Induced myeloid leukemia	A A12 700		5UUM*	6B4L*	
	1 A aggetin	0.998	-8.081		
differentiation protein MCL-1	1- Acacetin			-8.386	
"It is involved in regulation of	2- Ellagic acid	0.999	-8.125	-8.204	
apoptosis and cell survival" [16]	3- Kaempferol	1.0	-8.208	- 8.629	
	4- Quercetin	1.0	-8.369	- 8.047	
	a- Obatoclax	-	-8.357	-7.452	
	• A CJY 401	-	_	-9.461	
21- Serine/threonine-protein kinase pim-1			6AYD*	6BSK*	
"It is involved in tumorigenesis,	1- Acacetin	0.994	-7.992	-6.878	
cell survival, and proliferation" [16]	2- Ellagic acid	0.9	-7.155	-7.953	
	3- Kaempferol	0.99	-8.282	-7.732	
	4- Quercetin	1.0	-8.875	-8.342	
	a- Leucettine L41	_	-8.785	-8.134	
	• A C2J 401	_	-5.852	_	
	• A MVG 405	_	_	-8.970	
22- Matrix metalloproteinase 9	1111110 100		6ESM*	5CUH*	
"involved in tumor transformation,	1- Quercetin	1.0	- 11.134	-9.847	
progression, survival, angiogenesis	• A B9Z 306	1.0	-8.173	-	
and metastasis" [37]	• A LTQ 306	_	- 0.173	- - 11.035	
	ALIQ 500	_	- 4WH7*		
23- M phase inducer phosphatase	1 (1) M-11::4:	0.047		4WH9*	
"It is a key cell cycle regulator" [38]	1- (+)-Mollisacacidin	0.947	- 9.204	-8.835	
	2- Catechin	0.947	- 8.297	-8.199	
	3- Digallic acid	0.982	-10.300	- 8.991	
	4- Epicatechin	0.947	- 8.049	- 9.179	
	5- Kaempferol	0.994	-7.152	- 7.524	
	6- γ-Sitosterol	1.0	-7.169	-10.413	
	7- Niloticane	1.0	-7.099	-7.917	
	• A 8H8 607	-	-4.390	_	
	• A 3M8 601	_	_	-6.086	
24- Nuclear receptor ROR-alpha			1N83*	3B0W*	
"It is involved in cell growth, differentiation,	1- Acacetin	1.0	-8.342	-8.021	
and control of metastatic behavior of	2- Ellagic acid	0.969	-8.433	-8.275	
androgen-independent prostate cancer" [39]	3- Quercetin	0.857	-8.974	-9.179	
	• A CLR 1000	_	-11.358	_	
	• B DGX 1	_	_	-15.907	
25- Serine/threonine protein kinase Nek2	D D GAT I		2XNN*	2WQO*	
"It regulates centrosome separation, bipolar	1- Quercetin	1.0	- 8.417	-9.318	
spindle formation in cell mitosis" [16]	• A 430 1280	1.0	- 7.282		
spindle formation in cen fintosis [10]		_		10.022	
06 P. 1	• A VGK 1280	_	-	- 10.033	
26- P-glycoprotein 1 and 3			4XWK*	2CBZ*	
"They involved in multi-drug	1- Acacetin	1.0	-8.064	-8.580	
resistance" [40]	2- Chalconaringnen-4-O	0.979	-11.266	-9.360	
	3- Kaempferol	1.0	-7.418	-6.493	
	4- Kaempferol-7-glucoside	0.97	-11.300	-9.221	
	5- Querecitin	1.0	-8.002	-5.888	
	• A 4C8 1301	-	-8.874	_	
	• A ATP 1873	_	_	-9.674	
27- Platelet-derived growth factor 1 receptor			5GRN*	5K5X*	
"It has a pro-angiogenic action [41]	1- Ellagic acid	1.0	-8.681	- 7.922	
L D D D	a- Sunitinib	_	-9.363	-8.234	
28- Proto-oncogene tyrosine	a Samano		4MXO*	4MXY*	
protein kinase Src	1- Ellagic acid	1.0	- 7.488	- 7.260	
protein kinase sie	• B DB8 601	1.0	- 7.488 - 8.930	- 7.260 - 7.870	



Table 1 (continued)

Target	Compounds	LBVS	Docking score	
"It participates in cancer cells		,		
invasion and progression" [42]				
29- Protein kinase C epsilon type			1GMI*	2WH0*
"It is essential in cell invasion, adhesion,	1- Kaempferol	0.787	-7.836	-6.832
migration, and regulation of apoptosis" [16]	2- Naringenin	0.99	- 7.543	-6.772
30- Steroid 17 alpha-hydroxylase			6CIR*	5UYS*
"It is a key regulatory enzyme,	1- (+)-Mollisacacidin	0.83	-8.744	- 8.462
essential in androgens biosynthesis" [43]	2- Ellagic acid	1.0	-8.526	- 8.393
	$3-\gamma$ -Sitosterol	1.0	- 10.868	- 10.792
	4- Epicatechin	0.838	-9.273	-9.793
	5- Kaempferol	0.955	-8.360	- 8.455
	6- Niloticane	0.998	-8.485	-8.727
	7- Querecitin	1.0	-8.940	- 8.497
	a- Galeterone	_	-10.681	- 11.190
	• A 3NQ 601	_	- 10.485	- 10.777
21 T 1 1 12	• A 8QD 601	_	— 4T I C A *	- 10.777
31- Tankayrase enzyme 1 and 2	1 A	0.74	4U6A*	4HKI*
"Involved in cell cycle progression and	1- Acacetin	0.74	- 10.054	- 11.596
telomere homeostasis" [44]	2- Flavone	1.0	-9.475	- 9.961
	• A 3DN 1402	_	- 11.568	- 10.225
22 TI 1	• A FLN 1204	_		- 10.237
32- Telomerase reverse transcriptase enzyme	1 / > 3 / 11	0.001	5UGW*	5NPT*
"It is involved in the regulation of transcription	1- (+)-Mollisacacidin	0.991	- 6.778	-7.634
and has a major role in the activation of	2- Acacetin	1.0	- 5.023	-7.100
telomerase at cancer" [45]	3- Ellagic acid	1.0	- 5.547	-8.204
	4- Catechin	0.991	-6.668	-7.590
	5- Epicatechin	0.991	- 6.774	-7.570
	6- Kaempferol	1.0	- 5.993	-7.128
	7- Leucocyanidin	0.912	- 7.246	- 10.797
	8- Naringenin	1.0	- 5.848	- 7.543
	9- Quercetin	1.0	-5.720	-7.434
	a- Berberine	_	- 22.039	-9.507
20 T	• A GSH 1201	_	-7.916	-
33- Transcription factor p65	4 () 3 () 11	0.00=	2RAM*	5 U01*
"It promotes tumor cells proliferation,	1- (+)-Mollisacacidin	0.987	-8.153	-8.549
suppresses the apoptosis, attracts the	2- (+)-Catechin-5,7 digllate	0.998	- 11.618	- 11.996
angiogenesis, metastasis, remodels the	3- Acacetin	1.0	-6.599	- 7.443
local metabolism and energizes the	4- Catechin-7-O-gallate	0.883	- 8.719	-9.457
immune system to	5- Ellagic acid	0.994	-7.309	-7.547
favor tumor growth" [46]	6- Epicatechin	0.987	- 8.661	-9.370
	7- Kaempferol	1.0	- 7.535	-7.202
	8- Leucocyanidin	0.986	- 9.687	-9.905
	9- Quercetin	1.0	-7.730	-8.485
24 T	a- Bortezomib	_	- 6.232	-6.396
34- Tyrosine-protein kinase Lyn	1 1211 ' '1	1.0	5XY1*	3A4O*
"It is involved in the control of	1- Ellagic acid	1.0	-7.446	-8.238
proliferation and the inhibition of	• A 8H0 601	_	-8.054	-
apoptosis" [47]	• X STU 902	_	- 200/7D*	- 10.207
35- Vascular endothelial growth factor receptor 3	1.0	1.0	3WZD*	5EW3*
"It promotes tumor angiogenesis" [48]	1- Quercetin	1.0	- 8.561	-8.602
	a- Axitinib	_	- 8.699	-11.087
	b- Cabozantinib	_	-6.576	-7.936
	• A LEV 1201	_	-8.648	-
	• A 5T2 1201	_	_	-9.156

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Table 2 Ligand-based virtual screening and molecular docking results regarding the antibacterial activity

Target	Compounds	LBVS	Docking Sco	ore	
1- 3-oxyacyl-[acyl-carrier protein]			4BNT*	5OVK*	5END*
reductase FabG	1- Acacetin	0.68	-9.017	-7.925	-7.891
"It catalyzes the first reductive step in the elongation cycle of fatty	2- Catechin	0.60	- 10.605	-9.857	-8.487
acid biosynthesis" [26]	3- Epicatechin	0.60	- 10.538	-8.816	-8.518
,	4- Kaempferol	1.0	-9.184	-8.071	- 7.497
	5- Quercetin	1.0	-10.068	-8.324	-8.186
	6- C 36E 1248	_	-5.612	_	_
	7- D NDP 301	_	_	-14.254	_
2- Enoyl-acyl carrier protein reductase			4M87*	4NR0*	4O1M*
"It is a limiting step enzyme in	1- (-)-Epigallocatchin-7-gallate	0.43	-11.920	- 12.991	-13.66
fatty acid biosynthesis, has no homolog in mammals" [50]	2- (+)-Catechin-4, 5,digallate	0.55	- 13.771	-13.034	- 14.25
iii iiiaiiiiiais [30]	3- (+)-Catechin-5-gallate	0.6	- 12.496	-11.710	-13.05
	4- Quercetin	1.0	-10.248	-10.823	-10.12
	a- Isonazid	_	-5.757	-5.210	-5.594
	b- Triclosan	_	-7.254	-7.110	-7.585
	• `A NAD 301	_	-13.553	-13.298	-11.81
3- D-alanine D-alanine ligase enzyme			6DGI*	5C1P*	3R23*
"It is an essential bacterial enzyme in	1- Quercetin	1.0	-6.267	-8.968	-7.719
peptidoglycan biosynthesis" [51]	a- Adenosine-5'-diphosphate	_	-3.430	-8.350	-5.561
	• A GOL 401	_	-5.112	_	_
	• D ADP 401	_	_	-7.924	_
4- AmpC Beta-lactamase enzyme			2HDQ*	2PU2*	2R9W*
"It is responsible for hydrolysis of	1- D-pinitol	0.876	-8.334	-7.930	-7.400
beta-lactams, with substrate specificity toward cephalosporins, has an important	2- Niloticane	0.992	-8.691	-8.713	-7.221
role in cephalosporins resistance" [16]	3- Quercetin	1.0	-9.481	-9.128	-8.917
	a- Clavulanic acid	_	-8.808	-7.732	-8.108
	• A C21 501	_	-6.937	_	_
	• B DK2 701	-	_	-9.133	_

In Compounds, the numbers 1, 2, 3, ... indicate *A. nilotica*'s phytochemical constituents, letters a, b, ... indicate positive controls, • indicates the cocrystallized ligands, and the italic emphasis indicates compounds with the higher scores. At Ligand-Based Virtual Screening Score (LBVS sco.), en dash (–) means that the compound was not screened. Asterisk (*) indicates the PDB ID

The Predicted Toxicity

According to the results, 1,6-di-O-galloyl-beta-D-glucose, ellagic acid, kaempferol, and quercetin were non-toxic as well as non-carcinogen (Table 9).

Drug-Likeness, Lead-Likeness, and Synthetic Accessibility Prediction

According to the results, (+)-Mollisacacidin, Acacetin, Catechin, Epicatechin, Kaempferol, Naringenin, Niloctane, and Quercetin were found to be the best lead and drug candidates with good synthetic accessibility, followed by Digallic acid, Ellagic acid, Leucocyanidin, and Melacacidin (Table 10).

Discussion

Despite the enormous conducted studies on the pharmacology activity of *A. nilotica*'s [1], the determination of the target that contribute to its activity and the understanding of the mechanism of action as well as to assess the pharmacokinetics, safety, and the drug-likeness probability are important issues that were not conducted yet. Such studies are required to bring the plant in the drug discovery pipeline so as to design a novel drug with broad-spectrum of therapeutic activity and safety.

To identify the targets, TargetNet web servers that utilize a QSAR model based on the chemogenomic data as a predictive algorithm [25] and Similarity Ensemble Search Tool [24] were used. To validate the predicted target from the web servers, a molecular docking study was performed using Cresset Flare software [23] that uses the Lead finder program [69] for



Table 3 Ligand-based virtual screening and molecular docking results regarding the antiviral and the antiplasmodial activity

Target	Compounds	LBVS	Docking score	
1- HIV integrase enzyme			3ZT4*	5KGX*
"Antiviral target acts on the essential step in	1- Acacetin	1.0	-8.771	-4.595
viral replication cycle via catalyzes of viral DNA integration into host DNA" [52]	2- Digallic acid	1.0	-11.163	-7.227
integration into nost DNA [52]	3- Ellagic acid	0.75	-9.509	-4.765
	4- Kampeferol	1.0	-8.889	-5.473
	5- Naringenin	0.76	-9.118	-5.388
	6- Querecitin	_	-9.897	-5.120
	• A 7SK 301	_		-4.705
	• A ZT2 1217	_		-8.030
2- Corona virus replicase polyprotein 1 ab			5NH0*	5N5Ov*
"Antiviral target involved in transcription and	1- Quercetin	1.0	-9.436	-9.848
replication of viral RNAs and interacts with host 40S ribosomal subunit leading to translation	• A 8X8 301	_	-7.612	_
inhibition" [26]	• A 8O5 401	_	_	-10.887
3- Toll-like receptor 9			5Y3L*	5ZLN*
"Antiviral and antibacterial target acts as innate	1- Acacetin	1.0	-7.024	-6.001
immune receptor acts in recognition of microbial DNA" [53]	2- Ellagic acid	0.999	- 7.956	-6.225
Inicional DIVA [33]	3- Epicatechin	0.819	-8.138	-6.796
	4- Kaempferol	1.0	-7.726	-6.369
	5- Leucocyanidin	0.992	-9.222	-8.689
	6- Naringenin	0.889	-7.664	-7.244
	7- Niloctne	1.0	-7.967	-6.212
	8- Quercetin	1.0	-7.481	-7.048
4- β-Hydroxy acyl-ACP dehydratase FabZ			3AZB*	3AZA*
"Antiplasmodial target involved in the fatty	1- (+) Mollisacacidin	0.6	-8.297	- 10.946
acid biosynthesis" [54]	2- Acacetin	0.67	-6.787	-8.454
	3- Catechin	0.60	- 9.251	- 11.096
	4- Epicatechin	0.60	-8.857	-11.216
	5- Kaempferol	0.78	-7.318	- 9.544
	6- Quercetin	1.0	- 10.906	-9.286
	• G KM1	_	- 7.913	_
	• B KM0 2	_	=	-6.603
5- Hydroxyacyl-[acyl-carrier-Protein] dehydratase			3ED0*	3CF9*
"Antiplasmodial and antibacterial target responsible	1- Quercetin	1.0	-10.032	-9.607
for fatty acid biosynthesis" [55]	• A EMO 163	_	-8.454	_
	• A AGI 161	_	_	-7.841
6- MO15-related protein kinase pfmrk enzyme			Raptor x	Phyre 2
"Antiplasmodial target that is a cyclin-dependent kinase enzyme plays a central role in the regulation of cell cycle" [56]	1- Acacetin	0.74	-8.125	- 9.482
7- M18 aspartyl aminopeptidase enzyme			4EME*	Phyre 2
"Antiplasmodial target involved in host	1- (+) Mollisacacidin	1.0	-10.319	-8.855
erythrocyte invasion and the degradation of host hemoglobin" [57]	2- Epicatechin	1.0	- 9.281	-9.895
or nost nomogroum [5/]	3- Querecitin	1.0	-10.223	-9.218

In Compounds, the numbers 1, 2, 3, ... indicate *A. nilotica*'s phytochemical constituents, letters a, b, ... indicate positive controls, • indicates the cocrystallized ligands, and the italic emphasis indicates compounds with the higher scores. At Ligand-Based Virtual Screening Score (LBVS sco.), en dash (–) means that the compound was not screened. Asterisk (*) indicates the PDB ID. Phyre2 and Raptor x means the 3D structure of target was modeled by Phyre2 [58] and RaptorX [59] web servers, respectively



Table 4 Ligand-based virtual screening and molecular docking results regarding the antidiabetic activity

Target	Compounds	LBVS	Docking Score	e
1- Insulin receptor			2W12*	40GA*
	1- Ellagic acid	1.0	- 9.239	-8.785
	a- Ceritinib	_	-8.150	-8.093
2- Glycogen phosphorylase (muscle)			2ZB2*	3CEJ*
"An important allosteric enzyme in	1- Quercetin	1.0	-11.15	- 9.523
carbohydrate metabolism" [16], "potential target in type 2	• A A46 850	_	-11.44	-
diabetes mellitus" [56]	• A AVF 833	_		-11.52
3- Sodium/glucose co-transporter 2			2XQ2*	3DH4*
"The interaction with this enzyme	1- 1,6-di-O-galloyl-beta-D-glucose	1.0	-12.05	- 12.05
inhibits the renal glucose reabsorption, leading to a reduction	2- Chalconaringnen-4-O-beta.	1.0	-10.53	- 10.53
in plasma glucose level" [60]	3- Niloticane	0.961	-7.634	-7.822
	4- Canagliflozin	_	- 11.11	-8.030
4- Aldose reductase enzyme			3RX4*	3V36*
"It is involved in the development	1- (+)-Mollisacacidin	1.0	-9.074	-11.20
of the secondary diabetic complications" [61]	2- Acacetin	0.74	-9.196	-10.17
complications [01]	3- Chalconaringnen-4-O-beta-glucose	0.74	-12.16	- 14.31
	4- Catechin	1.0	-9.581	- 11.74
	5- Dicatechin	0.959	-12.70	-10.04
	6- Ellagic acid	1.0	-8.425	-8.886
	7- Epicatecin	1.0	-9.513	- 11.73
	8- Kaempferol	1.0	-8.185	-10.86
	9- Kaempferol-7-glucoside	0.97	- 12.35	- 14.27
	10- Leucocyanidin	1.0	-11.45	- 14.21
	11- Melacacidin	1.0	-9.75	-12.90
	12- Naringenin	1.0	-9.618	-9.869
	13- Quercetin	1.0	-8.824	-10.81
	a- Epalrestat	_	-8.849	- 10.50
	• A SFI 317	_	-8.393	-8.61
5- Beta-secretase enzyme			5MXD*	4BEL*
"It is down-regulator of insulin	1- Ellagic acid	1.0	-7.493	-7.677
receptors amounts and signaling in the liver" [62]	2- Quercetin	1.0	-6.955	-7.408
	a- 5,5-Diphenyliminohydantoin	_	-4.885	- 5.654
	• A III 701	_	-5.569	-
	• A B3P 1399	_	_	-6.797

In Compounds, the numbers 1, 2, 3, ... indicate A. nilotica's phytochemical constituents, letters a, b, ... indicate positive controls, • indicates the cocrystallized ligands, and the italic emphasis indicates compounds with the higher scores. At Ligand-Based Virtual Screening Score (LBVS sco.), en dash (–) means that the compound was not screened. Asterisk (*) indicates the PDB ID

docking calculation. Moreover, pkCSM [13] and SwissADME web servers [12] were used to predict the pharmacokinetics (ADME: Absorption, Distribution, Metabolism, and Elimination), toxicity, and the drug-likeness probability.

The total predicted targets form the virtual screening with the highest probability that was validated by the molecular docking were 61 targets.

The interaction of Acacetin with the cell division control protein 42 homolog (CDC42) will prevent the oncogenic transformations. The inhibition of enzymes—anaplastic lymphoma kinase by Quercetin, cyclin-dependent kinases 1, 4,

and 6 by Ellagic acid and Acacetin, Aurora A and B by Ellagic acid and Quercetin, serine/threonine protein kinase Nek2 by Quercetin, proto-oncogene tyrosine-protein Kinase Src by Ellagic acid, tankyrase 1 and 2 by Acacetin as well as M phase inducer phosphatase by Digallic acid, Epicatechin, and Kaempferol—will prevent the cancer progression and development.

Moreover, the inhibition of the enzymes—cell division cycle-7-related protein kinase by Ellagic acid, serine/threonine-protein kinase pim-1 by Quercetin, Ellagic acid, and Kaempferol as well as DNA topoisomerase 1 by



Table 5 Ligand-based virtual screening and molecular docking results regarding the anti-inflammatory activity

Target	Compounds	LBVS	Docking sco	ore
1- Arachidonate 15-lipoxygenase.			4NRE*	2P0M*
"Has an important role in the	1- (–)-Epigallocatchin-7-gallate	1.0	-11.046	-12.03
immune and the inflammatory	2- (+)-Mollisacacidin	1.0	-7.845	-9.535
responses" [16].	3- Acacetin	1.0	-7.693	-9.283
• • •	4- Catechin	0.984	-7.864	-10.05
	5- Dicatechin	1.0	-9.003	- 11.19
	6- Ellagic acid	1.0	-6.923	-8.525
	7- Epicatechin	1.0	-8.210	- 9.662
	8- Kaempferol	1.0	-7.659	-9.058
	9- Leucocyanidin	1.0	-10.035	- 12.76
	10- Naringenin	1.0	-7.857	-8.760
	11- Quercetin	1.0	-7.865	-9.732
	a- Diethylcarbamazine	_	-7.511	-7.623
	• A C8E 702	_	-5.722	_
	• B RS7 841	_	_	-6.830
2- Cycloxygenase-2 enzyme (COX-2).			5IKQ*	5F1A*
"It generates the inflammatory	1- (+)-Mollisacacidin	0.801	-10.127	-8.246
mediator's Prostaglandins	2- Acacetin	1.0	- 8.944	-7.561
from the arachidonic acid" [62].	3- Ellagic acid	1.0	-7.932	- 7.969
	4- Kaempferol	1.0	-9.149	<i>−</i> 7.444
	5- Leucocyanidin	0.907	-8.121	-5.193
	6- Naringenin	0.991	-9.426	-8.345
	7- Querecitin	1.0	-9.281	- 7.759
	a- Diclofenac	_	-8.628	-7.131
	b- Indomethacin	_	-8.771	-7.162
	• A JMS 602	_	-8.632	_
	• A SAL 601	_	_	-5.384
 Macrophage migration inhibitory factor. 			5XEJ*	6CB5*
"It is a pro-inflammatory cytokine	1- (–)-Epigallocatchin-5, 7-gallate	1.0	-8.500	- 11.51
counteracts the anti-inflammatory	2- (+)-Catechin-5,7-digallate	1.0	-10.457	- 12.24
activity of glucocorticoids" [16].	3- (+)-Mollisacacidin	1.0	-6.527	-9.613
	4- Catechin	1.0	-6.836	-9.855
	5- Dicatechin	0.966	- 7.621	-8.843
	6- Epicatechin	1.0	-7.600	− 9.411
	7- Kaempferol	1.0	-5.332	-8.927
	8- Leucocyanidin	1.0	-7.623	- 9.649
	9- Naringenin	1.0	-6.853	-8.881
	10- Querecitin	1.0	-5.987	-9.257
	a- 3,4-Dihydroxycinnamic acid	_	-7.846	-10.35
	• A 6UV 204	_	-8.978	_
	• A EV7 201	_		- 11.79
4- Phospholipase A2 enzyme.			2B96*	50W8*
"It is responsible for the release of the	1- γ-Sitosterol	0.831	-8.764	-7.798
arachidonic acid from arachidonyl	2- Digallic acid	0.884	- 8.626	-8.788
phospholipids, thereby involved	3- Kaempferol	0.938	-8.563	- 7.556
in the initiation of the	4- Quercetin	1.0	- 8.655	-8.101
inflammatory response" [16].	a- Prostaglandin A2	_	- 7.640	-5.746
5 P	• A ANN 501	_	- 5.760	- (ID 67%
5- Receptor-interacting serine/	1 4 2	0.072	6FU5*	6HMX*
threonine protein kinase 2.	1- Acacetin	0.962	- 8.185	-6.108
"It is involved in the formation	2- Digallic acid	0.924	- 12.137	-9.105
of the productive inflammatory	3- Ellagic acid	1.0	- 8.111	-7.246
Response" [63].	4- Kaempferol	0.919	-8.138	-6.606
	5- Niloctne	1.0	- 10.128	-7.078
	6- Quercetin	0.906	-8.320	-6.902
2 TT 41 14 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1	• A E7N 400	_	-9.371	-
6- Xanthine dehydrogenase/ oxidase	1 / \ 3 6 11'	1.0	2E1Q*	3AM9*
enzyme. "It contributes to uric acid formation and generation of	1- (+)-Mollisacacidin	1.0	- 11.185	-11.260
reactive oxygen species" [16].	2- 1,6-di-O-galloyl-beta-D-glu.	1.0	- 14.648	- 13.042
	3- Acacetin	1.0	- 9.690	-9.847
	4- Catechin	1.0	- 11.260	91.380
	5- Digallic acid	1.0	- 10.059	- 10.049
	6- Ellagic acid	1.0	-9.337	-11.830
	7- Epicatechin	1.0	- 11.367	- 11.669
	8- Kaempferol	1.0	- 9.900	-10.184
	9- Leucocyanidin	1.0	- 10.999	- 11.794
	10- Quercetin	1.0	-11.177	-11.371
	a- Allopurinol	_	-6.680	-6.120
	• A MTE	_	-13.547	-13.141

In Compounds, the numbers 1, 2, 3, ... indicate *A. nilotica*'s phytochemical constituents, letters a, b, ... indicate positive controls, • indicates the cocrystallized ligands, and the italic emphasis indicates compounds with the higher scores. At Ligand-Based Virtual Screening Score (LBVS sco.), en dash (–) means that the compound was not screened. Asterisk (*) indicates the PDB ID



Table 6 Ligand-based virtual screening and molecular docking results regarding the antidiarrheal, anti-platelets, and anticholinesterase activity

Target	Compounds	LBVS	Docking Score		
1- Mu-type opioid receptor			4DKL*	5C1M*	
"Antidiarrheal target, the action	1- (+)-Mollisacacidin	0.994	- 7.279	-9.051	
on mu and delta opioid receptors leads to inhibition of diarrhea	2- Catechin	0.994	-7.822	-8.727	
without constipation [64]	3- Dicatechin	0.983	-8.880	-12.268	
	4- Epicatechin	0.994	7.392	-8.836	
	a- Eluxadoline	_	-6.868	-10.799	
	• A 4VO 401	_	_	-11.228	
2- Delta-type opioid receptor			4N6H*	4EJ4*	
Antidiarrheal target	1- (+)-Catechin-3, 5-digallate	0.998	-11.44	-10.808	
	2- Dicatechin	0.999	-12.06	-11.488	
	a- Eluxadoline	_	-7.834	-7.712	
	• A EJ4 1219	_	-11.01	_	
	• A EJ4 500	_	_	-10.392	
3- P2Y12 receptor			4PXZ*	4NTJ*	
"Anti-platelets target has a	1- 1, 6-di-O-galloyl-beta-D-glucose	0.756	- 11.99	-11.397	
central role in platelet activation" [65]	2- Clopidogrel	_	-5.521	-6.873	
	• A 6AD 1201	_	-20.29	_	
	• A AZJ 1201	_	-19.78	_	
4- Acetylcholinesterase enzyme			1H22*	1ODC*	
"The reversible inhibition of	1- (+)-Catechin-4,5-digallate	0.73	-12.837	-13.137	
enzyme acetylcholinesterase increases the concentration of	2- (+)-Mollisacacidin	0.984	-10.697	-11.074	
acetylcholine that is helpful in	3- Acacetin	0.74	-8.728	-10.147	
neurodegenerative disorders like	4- Catechin	0.984	-11.143	-11.397	
Alzheimer's disease [66]	5- Digallic acid	0.94	-12.683	-13.272	
	6- Epicatechin	0.984	- 11.053	-10.701	
	7- Flavone	1.0	-7.519	-7.521	
	8- Leucocyanidin	0.99	- 12.967	-13.153	
	9- Lupenone	1.0	-10.976	-8.968	
	10- Lupeol	0.99	-11.737	-10.071	
	11- Melacacidin	0.99	-11.162	-11.291	
	12- Niloticane	1.0	-10.396	-10.735	
	a- Neostigmine	_	- 7.919	-8.196	
	• A E10 1536	_	-11.084	-11.266	
	• A A8B 1538	_	_	-11.266	

In Compounds, the numbers 1, 2, 3, ... indicate *A. nilotica*'s phytochemical constituents, letters a, b, ... indicates positive controls, • indicates the cocrystallized ligands, and the italic emphasis indicates compounds with the higher scores. At Ligand-Based Virtual Screening Score (LBVS sco.), en dash (–) means that the compound was not screened. Asterisk (*) indicates the PDB ID

Kaempferol-7-glucoside and 1,6-di-O-galloyl-beta-D-glucose—will suppress the DNA replication; the inhibition of enzymes—cyclin-dependent kinase 9 by Acacetin and telomerase reverse transcriptase by Leucocyanidin, Quercetin, Ellagic acid, and Kaempferol—will suppress the transcription; the inhibition of tyrosine-protein kinase Lyn enzyme by Ellagic acid will prevent the cancer cells proliferation; as well as the inhibition of angiopoietin-1 receptor, proto-oncogene tyrosine-protein kinase Src by Ellagic acid, and protein kinase C epsilon by kaempferol and Naringnen will suppress the cancer cells invasion.

Furthermore, the inhibition of ephrin type B receptor 4 and platelet-derived growth factor 1 receptor by Ellagic acid, vascular endothelial growth factor receptor 3 by Quercetin, as well as focal adhesion kinase enzyme Ellagic acid and Quercetin will suppress the angiogenesis.

The inhibition of P-glycoprotein 1, 3 transporters by Kaempferol-7-glucoside, Chalconaringnen-4-O-beta-glucopyranoside, Kaempferol, and Quercetin as well as ATP binding cassette sub-family G member 2 by (+)-Catechin-3, 5-digallate, Chalconaringnen-4-O-beta-glucopyranoside,



Table 7 The predicted pharmacokinetics properties of phytochemical constituents having higher affinity scores (part A)

Phytochemical constituent	Intestinal absorption	BBB permeability	Human Vd (L/kg)	Total clearance (mg/kg/day)	Renal OCT2 substrate
(-)-Epigallocatechin-5,7-gallate	Low (14.341%)	No	1.29	0.23	No
(-)-Epigallocatechin-7-gallate	Low (47.214%)	No	1.46	3.4	No
(+)-Catechin-3,5,-digallate	Low (44.42%)	No	0.97	0.3	No
(+)-Mollisacacidin	High (72.264%)	No	1.33	1.85	No
1,6-di-O-galloyl-beta-D-glucose	Low (28.679%)	No	2.65	5.4	No
Acacetin	High (94.546%)	No	0.78	5.9	No
Acanilol-1	High (98.82%)	No	0.34	5.5	No
Acanilol-2	High (96.44%)	No	0.3	4.6	No
Catechin	High (72.264%)	No	1.33	1.9	No
Catechin-7-O-gallate	(54.376%)	No	1.40	1.0	No
Chalconaringnen-4-O-beta-glucopyranoside	Low (17.445%)	No	0.51	1.79	No
Dicatechin	(69.966%)	No	0.75	1.73	No
Digallic acid	Low (47.548%)	No	1.26	3.5	No
Ellagic acid	High (76.935%)	No	0.97	4.1	No
Epicatechin	High (72.264%)	No	1.33	1.85	No
Flavone	High (94.935%)	Yes	0.96	2.12	No
γ-Sitosterol	High (95.884%)	No	1.32	4.25	No
Kaempferol	High (75.481)%	No	1.02	4.52	No
Kaempferol-7-glucoside	Low (44.274%)	No	0.67	5.17	No
Leucocyanadin	(65.231%)	No	2.8	1.47	No
Lupenone	High (100%)	No	1.1	1.27	No
Lupeol	High (98.249%)	No	0.7	4.19	No
Melacacidin	(67.928%)	No	4.7	1.01	No
Naringnen	High (89.345%)	No	0.53	1.33	No
Naringnen-7-O-beta-glucopyranoside	(56.167%)	No	1.3	1.74	No
Nilobamate	High (88.608%)	Yes	1.5	21.29	No
Niloctane	High (95.567%)	Yes	1.1	7.17	Yes
Quercetin	High (75.36%)	No	2.2	3.83	No

The italic emphasis indicates desirable prosperity

BBB blood-brain barrier, Vd volume of distribution, Renal OCT2 human organic cation transporter 2 [68]

Leucocyanidin, Quercetin, Ellagic acid, and Kaempferol will suppress the cancer cells resistance.

The interaction of Lupeol, Quercetin, Ellagic acid, and Kaempferol with the enzyme inducible nitric oxide synthase on macrophage will promote a tumoricidal action. The inhibition of the enzymes—Bcl-2-related protein A1 by Leucocyanidin, Quercetin, Ellagic acid, and Kaempferol, the induced myeloid leukemia differentiation protein MCL-1 by Acacetin, Quercetin, Ellagic acid, and Kaempferol as well as the interaction with enzymes: caspase 9 by (–)-Epigallocatechin-7-gallate, Quercetin, Ellagic acid, and Kaempferol, death-associated protein kinase 1 by Kaempferol and Quercetin—will induce cancer cell apoptosis. Consequently, those compounds show substantial anticancer activity (Table 1).

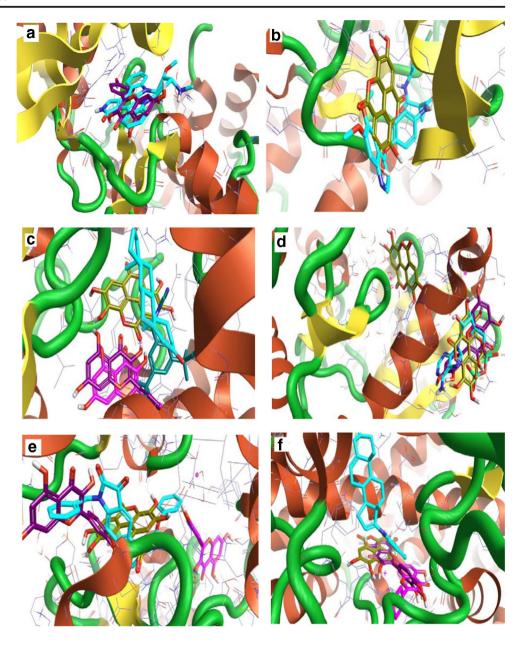
The interaction of Kaempferol and Quercetin with the enzymes 3-oxyacyl-[acyl carrier protein] reductase FabG

and the interaction of (-)-Epigallocatechin-7-gallate, (+)-Catechin-4, 5-digallate, and Quercetin with enoyl-acyl carrier protein reductase will inhibit the bacterial fatty acids biosynthesis that is essential in the formation of bacterial membrane phospholipids [70] leading to "an impairment in the cellular envelope structure and function, the ability to form biofilms as well as increasing the susceptibility to the environmental stress" [71]. Moreover, the inhibition of the enzyme D-alanine D-alanine ligase by Quercetin will suppress the peptidoglycans biosynthesis that is vital in bacterial cell structure causing loss of bacterial cell integrity [72]. Therefore, those compounds have significant antibacterial activity (Table 2).

The interaction of Leucocyanidin, Ellagic acid, Kaempferol, and Quercetin with Toll-like receptor 9 will activate this innate immune receptor that helps in the recognition of microbial DNA [53]. The interaction of Digallic acid,



Fig. 2 The 3D interaction between the best compounds with some of their predicted anticancer targets. a Ouercetin (violet) with anaplastic lymphoma kinase enzyme. Staurosporine (turquoise) as control. b Ellagic acid (dark yellow) with angiopoietin 1 receptor. Cabozantinib (turquoise) as control. c Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with the aromatase enzyme. Anastrozole (teal) and the co-crystallized ligand A ASD 601(turquoise) as a control. d Ellagic acid (dark yellow) and quercetin (violet) with Aurora A kinase enzyme. The cocrystallized ligand A ADP 401(turquoise) as a control. e Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with caspase 9 enzyme. The Isatin sulfonamide 34 (turquoise) as a control. Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with steroid 17 alpha-hydroxylase enzyme. Galeterone (turquoise) as a control



Acacetin, Ellagic acid, Kaempferol, and Quercetin with HIV integrase enzyme will inhibit the viral DNA integration into host DNA leading to the suppression of replication cycle [52]. The interaction of Quercetin with Coronavirus replicase polyprotein 1 ab will inhibit the transcription and replication of viral RNAs [26]. Thus, those *A. nilotica*'s phytochemical constituents exhibit considerable antiviral activity (Table 3).

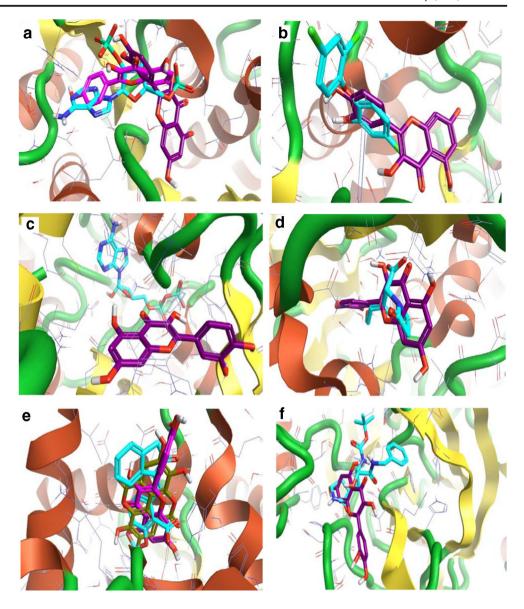
The interaction of Acacetin with the enzyme MO15-related protein kinase pfmrk will disrupt the regulation of plasmodial cell cycle [56], and the interaction of Quercetin and (+)-Mollisacacidin with the enzyme M18 aspartyl aminopeptidase will prevent the invasion in host erythrocyte and the degradation of host hemoglobin [57]. Furthermore, the interaction of Quercetin with the plasmodial enzymes β-hydroxy acyl-ACP dehydratase FabZ and hydroxyacyl-[acyl-carrier-Protein]

dehydratase will inhibit the fatty acid biosynthesis [54, 55] that are important for plasmodial membrane [73]. Subsequently, those *A. nilotica*'s phytochemical constituents show considerable antiplasmodial activity (Table 3).

The interaction of Ellagic acid with the insulin receptor will promote glucose uptake that lowers the blood glucose level [74]. The interaction of Quercetin with the enzyme glycogen phosphorylase will inhibit the glycogenolysis that reduces the hyperglycemia [75]. Moreover, the interaction of 1,6-di-Ogalloyl-beta-D-glucose and Chalconaringnen-4-O-beta-glucopyranoside with the sodium/glucose co-transporter 2 will inhibit the renal glucose reabsorption leading to a reduction in plasma glucose level [60], the interaction of Dicatechin, Kaempferol-7-glucoside, Leucocyanidin, Ellagic acid, Kaempferol, and Quercetin with the aldose reductase



Fig. 3 The 3D interaction between the best compounds with their predicted antibacterial and antiviral targets. a Kaempferol (pink), and quercetin (violet) with 3-oxyacyl-[acyl-carrier protein] reductase FabG. The cocrystallized ligand D NDP 301 (turquoise) as a control. b Quercetin (violet) with enoyl-acyl carrier protein reductase. Triclosan (turquoise) as a control. c Quercetin (violet) with Dalanine D-alanine ligase enzyme. The co-crystallized ligand D ADP 401(turquoise) as a control. **d** Quercetin (violet) with AmpC beta-lactamase enzyme. Clavulanic acid (turquoise) as a control. e Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with HIV integrase enzyme. The cocrystallized ligand A ZT2 1217 (turquoise) as a control. f Ouercetin (violet) with corona virus replicase polyprotein 1 ab. The co-crystallized ligand A 8X8 301 (turquoise) as a control



enzyme will suppress the development of the secondary Diabetic complications [61], as well as the interaction of Ellagic acid and Quercetin with the beta-secretase enzyme will upregulate the insulin receptors in the liver [62]. Successively, those *A. nilotica*'s phytochemical constituents have significant antidiabetic activity (Table 4).

The interaction of (-)-Epigallocatechin-7-gallate, Ellagic acid, Kaempferol, and Quercetin with the enzyme arachidonate 15-lipoxygenase and the interaction of Digallic acid, Acacetin, Ellagic acid, Kaempferol, and Quercetin with the receptor-interacting serine/threonine protein kinase 2 will interrupt the inflammatory responses [16, 63].

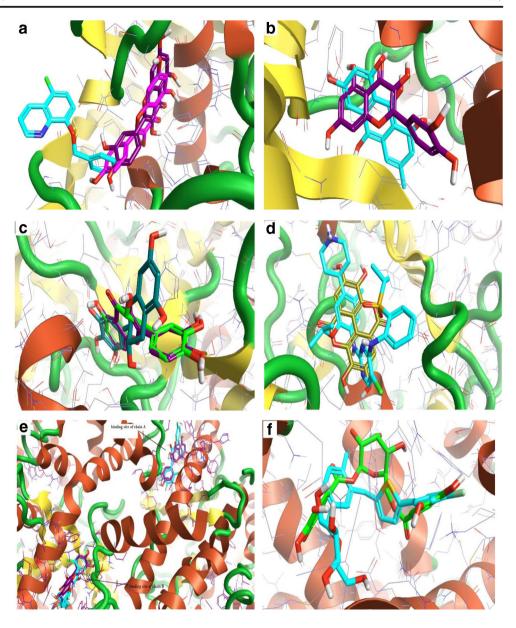
The interaction of (+)-Mollisacacidin, Naringnen, Ellagic acid, Kaempferol, and Quercetin with Cycloxygenase-2 enzyme (COX-2) will prevent the formation of inflammatory mediators prostaglandins [62], and

the interaction of Digallic acid, Kaempferol, and Quercetin with Phospholipase A2 enzyme will prevent the initiation of the inflammatory response [16]. Moreover, the interaction of 1, 6-di-O-galloyl-beta-D-glucose, Digallic acid, Acacetin, Ellagic acid, Kaempferol, and Quercetin with the xanthine dehydrogenase/oxidase enzyme will inhibit the formation of uric acid and reactive oxygen species [16]. Thus, those of *A. nilotica*'s phytochemical constituents exhibit considerable anti-inflammatory activity (Table 5).

The interaction of Dicatechin with the mu and delta opioid receptors will lead to antisecretory and anti-transit action that will inhibit diarrhea [76]. The interaction of 1,6-di-O-galloylbeta-D-glucose with P2Y12 receptor will inhibit the platelet activation [65]; consequently, they have considerable antidiarrheal and anti-platelets activity, respectively (Table 6).



Fig. 4 The 3D interaction between the best compounds with some of their predicted antiplasmodial and antidiabetic targets. a Kaempferol (pink), and quercetin (violet) with β-hydroxy acyl-ACP dehydratase FabZ. The co-crystallized ligand B KM0 2 (turquoise) as a control. b Quercetin (violet) with hydroxyacyl-[acyl-carrier-Protein] dehydratase. The cocrystallized ligand A EMO 163 (turquoise) as a control. c (+)-Mollisacacidin (green), epicatechin (teal), and quercetin (violet) with M18 aspartyl aminopeptidase enzyme. d Ellagic acid (dark yellow) with insulin receptor. Ceritinib (turquoise) as a control. e Quercetin (violet) with glycogen phosphorylase (muscle). The co-crystallized ligand AVF 833 (turquoise) as a control. f 1,6-di-O-galloyl-beta-D-glucose (green) with sodium /glucose cotransporter 2. Canagliflozin (turquoise) as a control



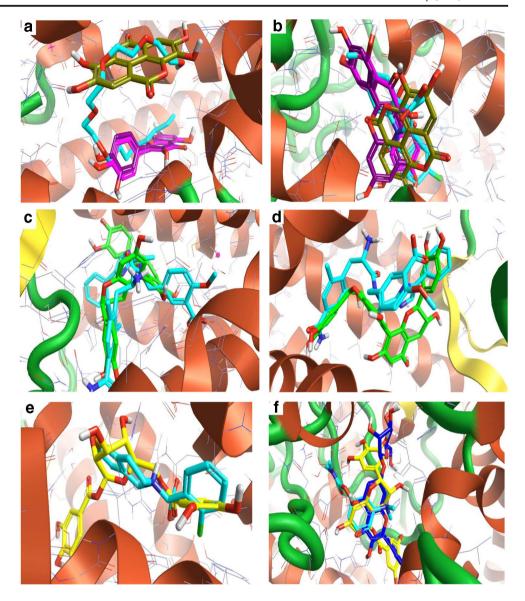
The chemogenomic-based QSAR models of TargetNet web server were strictly evaluated and validated leading to respected screening results [25]. Furthermore, the lead finder program [69] on Cresset flare software [23] is characterized by the combination between the genetic algorithm and different optimization strategies leading to great efficiency, robustness, accuracy, and speed of calculations [69]. Musab Ibrahim et al. [77] found the results of a molecular docking study about novel synthesized COX enzyme inhibitors conducted in Cresset Flare software were aligned with results of the conducted in vivo study. Depending on that, the obtained results of the predicted targets could be with considerable accuracy.

Since the pharmacological activity does not depend only on the pharmacodynamic properties, but also on the pharmacokinetic properties. Moreover, as the drug safety, the assessment of drug-likeness probability, and the synthetic accessibility are important issues [78], the identification of the best *A. nilotica*'s phytochemical constituents will be attained by the assessment of those issues collectively.

The pharmacokinetics is concerning the study of the entrance, movement, changing, and leaving of the drug to the body [79]. The higher absorption from the gastro-intestinal tract leads to higher drug concentration on the blood, the higher volume of distribution provides higher supply to the body tissues, and the adequate metabolism and elimination prevent the accumulation of the drug in the body, hence reduce the toxicity [79]. Consequently, the consideration of the pharmacokinetics in drug design is an essential task [80].



Fig. 5 The 3D interaction between the best compounds with some of their predicted antiinflammatory, antidiarrheal, and anti-platelets targets and acetyl cholinesterase enzyme. a Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with arachidonate 15-lipoxygenase enzyme. The co-crystallized ligand A C8E 702 (turquoise) as a control. b Ellagic acid (dark yellow), kaempferol (pink), and quercetin (violet) with cyclooxygenase-2 enzyme. Indomethacin (turquoise) as a control. c Dicatechin with Mu-type opioid receptor. Eluxadoline (turquoise) as a control. d Dicatechin with Delta-type opioid receptor. Eluxadoline (turquoise) as a control. e 1,6-di-O-galloyl-beta-Dglucose (yellow) with P2Y12 receptor. Clopidogrel (turquoise) as a control. f Digallic acid (yellow) and leucocyanidin (blue) with acetylcholinesterase enzyme. Neostigmine (turquoise) as a control



For instance, Lupenone is highly lipophilic; hence, it has higher absorption percent (100%); in contrast, the hydrophilic groups of Ellagic acid reduce it absorption percent to 76.935%, however, still it as a high absorption percent. The higher absorption will make Lupenone is highly bioavailable. Niloctane is permeable to BBB; therefore, its concentration that reaches the brain targets is more than Ellagic acid that is not permeable to the BBB. The predicted volume of distribution (Vd) of Melacacidin (4.7 L/kg) is the highest one, meaning that it has the highest distribution in body tissues. In contrary, Nilobamate has the highest predicted total clearance, meaning that it is the fastest one that eliminated from the body (Table 7).

Moreover, the inhibition of cytochrome-P enzyme CYP1A2 by Ellagic acid will decrease the biotransformation of drugs that metabolized by it leading to increase in the concentration of them that may increase the side effects;

consequently, the drug-drug interaction must be in consideration. The binding of Dicatechin with the P-glycoprotein may decrease the transportation of drugs transported by this transporter and may involve in the drug resistance by the pumping out mechanism (Table 8).

Furthermore, the predicted AMES toxicity of Epicatechin will lead to genotoxicity and mutagenicity [81], the predicted hERG II potassium channel inhibitory effect of Acacetin "prolongs the QT interval in ECG that increases the risk for potentially fatal ventricular arrhythmias" [82]; subsequently, such drugs will not be considered as drug-likeness candidate (Table 9).

Besides, Quercetin has no violation in Lipinski rule of five; hence, it will a good candidate as an orally active drug as well as it has no violation in Ghose, Veber, and Egan filters; therefore, it will be a good lead-likeness candidate [12] (Table 10).



Table 8 The predicted pharmacokinetics properties of phytochemical constituents having higher affinity scores (part B)

Phytochemical constituent	CYP2D6 and CYP3A4 substrate	CYP enzymes inhibition	P-gp substrate	P-gp I or II inhibition
(-)-Epigallocatechin-5, 7-gallate	No	Non inhibitor	Substrate	P-gp II
(-)-Epigallocatechin-7-gallate	No	CYP1A2, CYP3A4	Substrate	Non inhibitor
(+)-Catechin-3,5,-digallate	No	CYP2C9	Substrate	P-gp II
(+)-Mollisacacidin	No	Non inhibitor	Substrate	Non inhibitor
1,6-di-O-galloyl-beta-D-glucose	No	Non inhibitor	Substrate	Non inhibitor
Acacetin	No	CYP1A2, CYP2C19, CYP2C9, CYP2D6 CYP3A4 inhibitor	Substrate	Non inhibitor
Catechin	No	Non inhibitor	Substrate	Non inhibitor
Catechin-7-O-gallate	No	CYP1A2 inhibitor	substrate	P-gp II
Chalconaringnen-4-O-beta-glucopyranoside	No	Non inhibitor	Substrate	Non inhibitor
Dicatechin	No	Non inhibitor	Substrate	P-gp I and II
Digallic acid	No	CYP1A2, CYP3A4 inhibitor	Substrate	Non inhibitor
Ellagic acid	No	CYP1A2 inhibitor	Substrate	Non inhibitor
Epicatechin	No	Non inhibitor	Substrate	Non inhibitor
Flavone	CYP3A4	CYP1A2, CYP2C19, CYP2C9, and CYP2D6	Substrate	P-gp II
γ-Sitosterol	CYP3A4	Non inhibitor	No	P-gp I and II
Kaempferol	No	CYP1A2 inhibitor	Substrate	Non inhibitor
Kaempferol-7-glucoside	No	Non inhibitor	Substrate	Non inhibitor
Leucocyanadin	No	Non inhibitor	Substrate	Non inhibitor
Lupenone	CYP3A4	Non inhibitor	No	P-gp I and II
Lupeol	CYP3A4	Non inhibitor	No	P-gp II
Melacacidin	No	Non inhibitor	Substrate	Non inhibitor
Naringenin	No	CYP1A2, CYP2C19, CYP3A4	Substrate	P-gp II
Naringenin-7-O-beta-glucose	No	Non inhibitor	Substrate	P-gp II
Nilobamate	No	Non inhibitor	Substrate	Non inhibitor
Niloctane	No	Non inhibitor	Substrate	Non inhibitor
Quercetin	No	CYP1A2 inhibitor	Substrate	Non inhibitor

The italic emphasis indicates desirable prosperity, the bold emphasis indicates undesirable prosperity *CYP* cytochrome-P enzyme, *P-gp* P-glycoprotein transporter

According to the results of pharmacodynamics, pharmacokinetics, safety, and drug-likeness predictions collectively, Ellagic acid, Kaempferol, and Quercetin were the best *A. nilotica*'s phytochemical constituents that contribute to the therapeutic activities. The 3D interaction with their predicted targets demonstrates marked ligand superimposing with the control compounds (e.g., Figs. 1a, b, 2e, and 5b); however, it may at the same active site without ligand superimposing (e.g., Figs. 1c and 2c). Ellagic acid interacts with Aurora A kinase enzyme with two binding modes (Fig. 1d). Ellagic acid, Kaempferol, and Quercetin interact with Steroid 17 alpha-hydroxylase enzyme at a binding mode that differs from the binding mode of control Galeterone (Fig. 1f).

They were followed by (+)-Mollisacacidin, Epicatechin, and Melacacidin, those of their predicted AMES toxicity

decreased their rank. The predicted hERG II potassium channel inhibitory effect of Acacetin decreased its rank; however, it has good pharmacodynamics and pharmacokinetics profile.

Despite the efficient pharmacodynamics and the respectable safety profile of Ellagic acid, Kaempferol, and Quercetin, practically, each compound suffers from the low bioavailability [83–85], albeit the predicted intestinal absorption of them is high (Table 7). The reduced bioavailability of Ellagic acid is attributed to the poor absorption and rapid elimination from the body [86] (the predicted total clearance of Ellagic acid is high). The higher topological polar surface area (TPSA) (Table 10) contributes to the poor absorption. The reduced absorption of Kaempferol is attributed to the larger particle size and poor water solubility [83]. The reduced bioavailability of Quercetin is attributed to "the poor solubility and crystalline form at body temperature" [85].



Table 9 The predicted toxicity of phytochemical constituents having higher affinity scores

Phytochemical constituent	AMES tox.	hERG I or II inhibition	Hepatotoxicity	Skin sensitization	Carcinogenicity	Human maximum tolerated dose (mg/kg/ day)	Oral rat acute toxicity (mol/ kg)	Oral rat chronic tox. (mg/kg_bw/ day)
(-)-Epigallocatechin-5,7-gallate	No	hERG II	No	No	No	2.73	2.507	6.693
(-)-Epigallocatechin-7-gallate	Yes	hERG II	No	No	No	4.12	2.728	4.827
(+)-Catechin-3,5,-digallate	No	hERG II	No	No	No	2.37	2.506	6.19
(+)-Mollisacacidin	Yes	No	No	No	No	2.57	2.057	1.881
1,6-di-O-galloyl-beta-D-glucose	No	No	No	No	No	11.83	2.736	4.203
Acacetin	No	hERG II	No	No	No	5.74	2.558	1.467
Acanilol-1	Yes	hERG II	Yes	No	No	1.62	2.52	1.055
Acanilol-2	Yes	hERG II	No	No	No	1.65	2.256	1.865
Catechin	Yes	No	No	No	No	2.57	2.057	1.881
Chalconaringnen-4-O-beta-glucopyranoside	Yes	No	No	No	No	6.76	2.633	3.95
Dicatechin	No	hERG II	No	No	No	1.729	2.463	4.729
Digallic acid	Yes	hERG II	No	No	No	3.86	2.744	4.743
Ellagic acid	No	No	No	No	No	5.77	2.401	2.013
Epicatechin	Yes	No	No	No	No	2.57	2.057	1.881
γ-Sitosterol	No	hERG II	No	No	No	0.46	2.854	1.085
Kaempferol	No	No	No	No	No	8.13	2.301	2.699
Kaempferol-7-glucoside	Yes	hERG II	No	No	No	7.64	2.468	4.273
Leucocyanadin	Yes	No	No	No	No	9.4	2.175	2.974
Lupenone	No	hERG II	No	No	No	3.48	2.353	1.015
Lupeol	No	No	No	No	No	0.135	2.712	1.688
Melacacidin	Yes	No	No	No	No	10.47	2.092	3.379
Naringnen	No	hERG II	No	No	No	2.36	2.132	1.995
Naringnen-7-O-beta-glucopyranoside	Yes	No	No	No	No	4.73	2.583	4.019
Niloctane	No	No	No	No	No	0.27	2.419	1.788
Quercetin	No	No	No	No	No	11.07	2.221	2.997

The italic emphasis indicates desirable prosperity, the bold emphasis indicates undesirable prosperity *AMES tox.* AMES toxicity

Moreover, Ellagic acid, Kaempferol, and Quercetin have many polar phenolic hydroxyl groups (structure l, o, and x); consequently, they are subjected to direct glucuronide conjugation with as phase II metabolism. "Kaempferol and Quercetin are rapidly excreted in urine as glucuronides mainly" [87].

The reduced bioavailability affects pharmacological activity. Hence, to maintain the pharmacological activity, the bioavailability must be enhanced. The nano-suspension form of Kaempferol is increased its absorption and bioavailability [83]. The administration of Isoquercetin (Quercetin-3-glucoside) increases the absorption and bioavailability of Quercetin [88]. The bioavailability Ellagic acid, Kaempferol, and Quercetin is increased bypassing the entero-hepatic phase II conjugation (e.g., formation of ester derivatives) and by using novel drug delivery systems

as the liposomes. Furthermore, the co-administration of *Ginkgo biloba* extract with Kaempferol and Quercetin increased the bioavailability of them [89]. Consequently, the combination of Ellagic acid, Kaempferol, and Quercetin will be optimum treatment choice that maximizes the therapeutic activity and the safety profile as well as overwhelms the limits in the bioavailability. As they naturally are available in one plant, the combination of them at the therapeutic doses will be additive and will not induce drugdrug interactions. The design of multi-target drug is an effective promising approach for the treatment of complex disease [90].

The computational methods including the virtual screening are not to substitute the in vitro and in vivo methods, however, to reduce the time, cost, and the difficultness in the drug target identification [91]. Therefore, this study is an attempt to



Table 10 The predicted lead likeness, drug likeness, and synthetic accessibility score of phytochemical constituents having higher affinity scores

lable 10 The predicted	lable 10 The predicted lead likeness, drug likeness, and synthetic accessibility score of phytochemical constituents having higher affinity scores	and synthetic accessibility s	core of phytochemica	l constituents having higi	ner affinity scores		
Phytochemical constituent	Lipinski rule of five ^a	Ghose filters ^a	Veber filters ^a	Egan filters ^a	Muegge filters ^a	Lead likeness	Synthetic accessibility
(-)-Epigallocatchin-5, 7-gallate	3 violations (MW > 500, rotatb. bonds	3 violations (MW > 480, M R > 130,	I violation (TPSA > 140)	I violation (TPSA > 131.6)	4 violations (MW > 600, TPSA > 150, H-acc > 10, H. Am. > 51	I violation (MW > 350)	5.84
(-)-Epigallocatchin- 7-gallate	2 violations (rotatb. bonds > 10, H-don > 5)	No violations	I violation (TPSA > 140)	I violation (TPSA > 131.6)	3 violations (TPSA > 150, H-acc > 10,	I violation $(MW > 350)$	4.07
(+)-Catechin- 5,7-digallate	3 violations (MW > 500, rotatb. bonds > 10 H-don > 5)	2 violations (MW > 480, M R > 130)	I violation (TPSA > 140)	I violation (TPSA > 131.6)	H-don > 5) 3 violations (TPSA > 150, H-acc > 10, H-don > 5)	I violation $(MW > 350)$	4.62
(+)-Mollisacacidin	Yes	Yes	Yes	Yes	Yes	Yes	3.50
1,6-di-O-galloyl- beta-D-glucose	I violation (H-don > 5)	I violation (WLOGP < -0.4)	I violation (TPSA > 140)	I violation (TPSA > 131.6)	2 violations (TPSA > 150,, H-don > 5)	Yes	4.17
Acacetin	Yes	Yes	Yes	Yes	Yes	Yes	2.98
Catechin	Yes	Yes	Yes	Yes	Yes	Yes	3.50
Catechin- 7-O-gallate	I violation (H-don > 5)	Yes	I violation (TPSA > 140)	I violation (TPSA > 131.6)	2 violations (TPSA > 150, H-don > 5)	$I \ violation$ (MW > 350)	4.03
Dicatechin	3 violations (MW > 500, H-acc > 10, H-don > 5)	2 violations (MW > 480, M R > 130)	I violation (TPSA > 140)	I violation (TPSA > 131.6)	3 violations (TPSA > 150, H-acc > 10, H-don > 5)	I violation (MW > 350)	5.32
Digallic acid	Yes	Yes	I violation (TPSA > 140)	I violation (TPSA > 131.6)	2 violations (TPSA > 150, H-don > 5)	Yes	2.45
Ellagic acid	Yes	Yes	I violation (TPSA > 140)	I violation (TPSA > 131.6)	Yes	Yes	3.17
Epicatechin	Yes	Yes	Yes	Yes	Yes	Yes	3.50
γ -Sitosterol	I violation (MLOGP > 4.15)	3 violations (WLOGP > 5.6, MR > 130, atoms > 70)	Yes	I violation (WLOGP > 5.88)	2 violations (XLOGP3 > 5, heteroatoms < 2)	I violation (MW > 350, XLOGP > 3.5)	6.30
Kaempferol	Yes	Yes	Yes	Yes	Yes	Yes	3.14
Kaempferol- 7-glucoside	2 violations (H-acc > 10, H-don > 5)	Yes	I violation (TPSA > 140)	I violation (TPSA > 131.6)	3 violations (TPSA > 150, H-acc > 10, H-don > 5).	I violation (MW > 350)	5.24
Leucocyanadin	Yes	Yes	Yes	Yes	I violation (H-don > 5).	Yes	3.76



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Phytochemical constituent	Lipinski rule of five ^a	Ghose filters ^a	Veber filters ^a	Egan filters ^a	Muegge filters ^a	Lead likeness	Synthetic accessibility
Melacacidin	I violation (H-don > 5).	Yes	Yes	Yes	I violation (H-don > 5).	Yes	3.75
Naringnen	Yes	Yes	Yes	Yes	Yes	Yes	3.01
Niloctane	Yes	Yes	Yes	Yes	Yes	Yes	5.37
Quercetin	Yes	Yes	Yes	Yes	Yes	Yes	3.23

The synthetic accessibility is from 1 (very easy) to 10 (very difficult). The bold indicates desirable prosperity, the Italic indicates undesirable prosperity as well as the bold-italic indicates the best compounds WW molecular weight, rotatb. bonds rotatable bonds, MR molar refractivity, H-don hydrogen bond donors, H-acc hydrogen bond acceptors, TPSA topological polar surface area

including molecular weight, number of parameter has a specific range that the structure must not be under physiochemical as well as the topological polar surface area. pharmacokinetic and between the the relationship otatable bonds, molar refractivity, number of hydrogen bond donors and describe and Muegge Veber, J bove it to be free from violation [12] ^aLipinski rule of five, Ghose,

identify the best A. nilotica's phytochemical constituents that contribute to its pharmacological activity as well as their targets. It is not meaning that this study alone will be sufficient to judge about the result; however, experimental studies are required to validate the results.

Conclusion

According to the results of pharmacodynamics, pharmacokinetics, safety, and drug-likeness predictions collectively, Ellagic acid, Kaempferol, and Quercetin were the best A. nilotica's phytochemical constituents that contribute to the therapeutic activities; consequently, we recommend the use of Ellagic acid, Kaempferol, and Quercetin as a combined drug via the novel drug delivery systems for the treatment of recent diseases attacking the public health including cancer, multidrug-resistant bacterial infections, diabetes mellitus, and chronic inflammatory systemic diseases. Moreover, we recommend wet lab studies to validate the results.

Compliance with Ethical Standards

Conflict of Interest The authors declare that having no conflict of interest.

Human and Animal Rights and Informed Consent This article does not contain any studies with human or animal subject performed by any of the authors

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