BUKTI KORESPONDENSI

ARTIKEL JURNAL INTERNASIONAL BEREPUTASI

Judul artikel	:	The Prediction of Pharmacokinetic Properties of Compounds in <i>Hemigraphis alternata</i> (Burm.F.) T. Ander Leaves Using pkCSM
Jurnal	• •	Indonesian Journal of Chemistry, 2022, Vol. 22(4), 1081-1089
Penulis Koresponden	:	Yeni Yeni

No.	Perihal	Tanggal
1	Bukti konfirmasi submit artikel dan artikel yang	18 Februari 2022
	disubmit	
2	Bukti konfirmasi review dan hasil review pertama	01 Maret 2022
3	Bukti konfirmasi submit revisi pertama, respon	13 April 2022
	kepada reviewer, dan artikel yang diresubmit	
4	Bukti konfirmasi review dan hasil review kedua	06 Mei 2022
5	Bukti konfirmasi submit revisi kedua, respon kepada	13 Mei 2022
	reviewer, dan artikel yang diresubmit	
6	Bukti konfirmasi artikel accepted	16 Mei 2022
7	Bukti tahap editing (copyediting, layout dan	16 Mei- 18 Juli 2022
	proofreading) artikel	
8	Bukti konfirmasi artikel published online	10 Agustus 2022

Bukti konfirmasi submit artikel dan artikel yang disubmit (18 Februari 2022)

Indonesian Journal of Chemistry

ISSN: 1411-9420

Published by Chemistry Department Universitas Gadjah



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Submission

Authors Yeni Yeni, Rizky Arcinthya Rachmania

Title The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander

Leaves Using pkCSM

73117-246938-1-SM.DOCX 2022-02-18 Original file

Supp. files None

Submitter Ms Yeni Yeni 🖾

Date submitted February 18, 2022 - 01:32 AM

Section Articles Stalis Ethica 🖾 Editor Abstract Views 10941

THE PREDICTION OF PHARMACOKINETIC PROPERTIES OF COMPOUNDS IN HEMIGRAPHIS ALTERNATA (BURM.F.) T. ANDER LEAVES USING PKCSM

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ABSTRACT

The inflammatory process has a role in the healing process and leads the body's homeostasis normal. Untreated acute inflammation can lead to organ pathology leading to a chronic inflammatory phenotype. Prediction of the affinity of 22 compounds in *Hemigraphis alternata* leaves as anti-inflammatory has been conducted for COX-1, COX-2 and 5-LOX. The Prediction of pharmacokinetic properties of the compounds was carried out to obtain inflammatory drug candidates that have high affinity with adequate ADME. The application used in this research is pkCSM, a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. The pkCSM used 20 predictors which were divided into 4 properties, absorption (7 predictors), distribution (4 predictors), metabolism (7 predictors) and excretion (2 predictors). Based on the prediction results, there are 5 compounds that have the best pharmacokinetic properties, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine. 3,7,11,15-Tetramethyl-2-hexadecen-1-ol is a compound that is predicted to have high affinity as an anti-inflammatory and good ADME.

Keywords: Pharmacokinetic, *Hemigraphis alternata*, pkCSM.

INTRODUCTION

Inflammation is a protective reaction of the immune system to defend the body from potentially harmful stimuli, both infectious and non-infectious agents that cause cell damage. It induces the inflammatory cells to be active and trigger inflammatory signaling pathways. The inflammatory process has an important function in the healing process so that abnormal body homeostasis can be restored. Acute inflammation that is not treated properly can cause organ pathology to worsen and eventually lead to a chronic inflammatory phenotype [1–3].

Hemigraphis alternata has various medicinal properties, including anti-inflammatory, antinociceptive and anti-diarrhoeal activities. Methanol and ethyl acetate extract of Hemigraphis alternata leaves has been shown to provide anti-inflammatory and non-toxic effects to mice [4]. There are 22 secondary metabolites that have been isolated in the leaves of this plant [5]. These compounds have predicted anti-inflammatory activity against COX-1, COX-2 and 5-LOX receptors. The compound predicted to have the highest affinity for COX-1 and COX-2 were 3,7,11,15-Tetramethyl-2-hexadecen-1-ol [6]. Meanwhile, the compound predicted to produce the highest affinity for 5-LOX was n-Hexadecanoic acid [7].

The relationship of pharmacokinetic properties, toxicity, and potency greatly affect the effectiveness of a drug. Determining the pharmacokinetic profile of a compound is carried out to determine the absorption, distribution, metabolism, and excretion (ADME) properties [8]. The initial assessment of ADME properties will help pharmaceutical researchers to select the best drug candidates for development and reject drug candidates that have a low probability of success. The aim of the in silico prediction of ADME properties is the accurate prediction of the in vivo pharmacokinetic properties of potential drug molecules in humans using only virtual structures [9]. The purpose of this study was to predict the pharmacokinetic properties (ADME) of 22 compounds contained in *Hemigraphis alternata* leaves through an in silico study.

EXPERIMENTAL SECTION

Provide sufficient detail to allow the work to be reproduced, including Materials, Instrumentation, and procedures.

Materials

The SMILES format of 22 compounds contained in the leaves of *Hemigraphis alternata* was obtained from PubChem (pubchem.ncbi.nlm.nih.gov). Compounds that do not have the SMILES format in PubChem can be obtained by describing their structure in the Online SMILES Translator (https://cactus.nci.nih.gov).

Instrumentation

The prediction of pharmacokinetic properties of 22 compounds in the leaves of *Hemigraphis alternata* was conducted using pkCSM (http://biosig.unimelb.edu.au/pkcsm/prediction).

Procedure

pkCSM is based on general properties of compounds (molecular properties, toxicophores and pharmacophores) and distance-based graph signatures. In pkCSM there are 20 predictors that describe the pharmacokinetic properties of a compound. The predictors were divided into absorption of 7 predictors, distribution of 4 predictors, metabolism of 7 predictors and excretion of 2 predictors (**Table 1**) [8,10,11].

Table 1. Distribution of ADME predictors in pkCSM

Property	Predictor	Unit	Property	Predictor	Unit
Absorption	Water solubility	log mol/L	Distribution	CNS	log PS
	(A1)			permeability	
				(D4)	
Absorption	Caco2	log Papp in	Metabolism	CYP2D6	Yes/No
	permeability (A2)	10-6 cm/s		substrate	
				(M1)	
Absorption	Intestinal	% Absorbed	Metabolism	CYP3A4	Yes/No
	absorption			substrate	
	(human) (A3)			(M2)	
Absorption	Skin Permeability	log Kp	Metabolism	CYP1A2	Yes/No
	(A4)			inhibitor	
				(M3)	
Absorption	P-glycoprotein	Yes/No	Metabolism	CYP2C19	Yes/No
	substrate (A5)			inhibitor	
				(M4)	
Absorption	P-glycoprotein I	Yes/No	Metabolism	CYP2C9	Yes/No
	inhibitor (A6)			inhibitor	
				(M5)	
Absorption	P-glycoprotein II	Yes/No	Metabolism	CYP2D6	Yes/No
	inhibitor (A7)			inhibitor	
				(M6)	
Distribution	VDss (human)	log L/kg	Metabolism	CYP3A4	Yes/No
	(D1)			inhibitor	
				(M7)	
Distribution	Fraction unbound	Fu	Excretion	Total	log
	(human) (D2)			Clearance	ml/min/kg
				(E1)	
Distribution	BBB permeability	log BB	Excretion	Renal	Yes/No
	(D3)			OCT2	
				substrate	
				(E2)	

In this study, virtual screening was carried out to obtain several compounds that had good ADME. Virtual screening is based on the results of the ADME predictor which has a numerical value with certain limitations. The predictors were Caco2 permeability (A2), intestinal absorption (human)

(A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3), CNS permeability (D4), total clearance (E1) [12]. The requirements are log Papp in 10-6 cm/s in Caco2 permeability (A2) > 0.9, intestinal absorption (human) (A3) > 30%, log Kp \geq -2.5 in skin permeability (A4), log L/kg \geq -0.15 in VDss (human) (D1), log BB \geq -1 in BBB permeability (D3), log PS \geq -3 in CNS permeability (D4).

The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54, 5 compounds with the best ADME were obtained. Subsequently, these results were compared with the predicted affinity in previous studies and 1 compound with good affinity and ADME was obtained [8].

RESULTS AND DISCUSSION

pkCSM is a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. pkCSM adapts the Cutoff Scanning concept to represent molecular and chemical structures to represent and predict their pharmacokinetic properties [8].

Water solubility is an important factor for a drug to show a better pharmacological response for oral administration. Drugs that have good water solubility properties will cause the drug to have good absorption and bioavailability properties. Good drug absorption and bioavailability can increase the plasma drug concentration at the target site to perform therapeutic functions [13].

Caco-2 cells are a cell line derived from colorectal adenocarcinoma [14]. The Caco-2 model was used to predict the possible gastrointestinal permeability of drugs in pre-clinical trials. This model expresses cytochrome P450 enzymes, transporters, microvilli, and enterocytes based on characteristics identical to those of the human small intestine [15]. The permeability of Caco-2 of a compound is high if it has a Papp > 8 x 10-6 cm/s. The permeability of Caco-2 is high if the predictive value is > 0.90 in the pkCSM predictive model [8].

The intestine is the main site for absorption of oral drugs. The compound is predicted to have low Intestinal absorption (human) if the value is < 30% [8]. The skin is the boundary between the internal and external body environment. Skin characteristics and properties can modify and affect drug delivery and toxicity [16]. A compound has a tendency to be skin permeable which is expressed by the skin permeability constant logKp (cm/hour). The logKp value of the compound > -2.5 indicates that the compound has low skin permeability [8].

Substrate P-glycoprotein is an ATP-binding cassette (ABC) transporter that functions as a biological barrier by removing toxins and xenobiotics from cells. P-glycoprotein I/II inhibitor is the ability of the compound to inhibit the transport of P-glycoprotein I and P-glycoprotein II. P-glycoprotein-mediated modulation of transport has significant pharmacokinetic implications for Pgp substrates. Inhibition of P-glycoprotein I or P-glycoprotein II can be exploited for certain therapeutic advantages or produce contraindications [8].

VDss (human) is the total amount of drug in the body divided by the total concentration of drug in plasma at steady state. This condition occurs when the system undergoes an infusion of the drug at a constant rate into the plasma and all drug concentrations in the body do not change [17]. VDss is low if the value is < 0.71 L/kg (log VDss < -0.15) and high if the value is > 2.81 L/kg (log VDss > 0.45). The effectiveness of a drug can be affected by the ability of the drug for binding proteins in the blood. The larger the fraction of the drug that is not bound to protein (fraction unbound), the more efficiently the drug will cross the cell membrane or diffuse [8].

Increased permeability of the BBB (Blood-Brain Barrier) which is a physical and biochemical barrier that plays a role in the defense of cerebral homeostasis can affect the pathological development of ischemic tissue [18,19]. A compound with a logBB > 0.3 can easily pass through BBB and a compound with a logBB < 1 is not well distributed to the brain. The ability of drug compounds to penetrate the CNS (Central Nervous System) can be determined from the value of the blood-brain permeability surface area product (logPS). This value was obtained from in situ brain perfusion with the compound directly injected into the carotid artery without any systemic distribution effect that could distort brain penetration. Compounds with logPS > -2 could penetrate the CNS, while compounds with logPS < -3 could not penetrate the CNS [8].

Cytochrome P450 is a detoxifying enzyme that can be found in the liver. Cytochrome P450 generally plays a role in drug metabolism. However, P450 inhibitors can dramatically alter drug pharmacokinetics. Therefore, it is important to know whether the given compound is a CYP2D6/CYP3A4 substrate predicted to be metabolized by P450. Cytochrome P450 oxidizes xenobiotics to be finally excreted. Many drugs are inactivated by cytochrome P450 and some can be activated by it. These enzyme inhibitors can affect drug metabolism and are contraindicated. It is therefore important to assess the ability of the compound to inhibit cytochrome P450 (isoform CYP1A2/CYP2C19/CYP2C9/CYP2D6/CYP3A4). A compound is considered a cytochrome P450 inhibitor if the concentration required to cause 50% inhibition is < 10 M [8].

Drug clearance is the volume of plasma in the vascular compartment that is cleared of drug per unit time. Total clearance is the sum of all body clearances. Total clearance gives an indication of drug elimination from the central compartment without reference to the mechanism of the process [20]. Organic Cation Transporter 2 (OCT2) is a renal uptake transporter that has an important role in the disposition and renal clearance of drugs or endogenous compounds. OCT2 substrates may cause adverse effects when administered concomitantly with OCT2 inhibitors [8].

In the results of the prediction of adsorption properties obtained compounds 3, 6, 9, 11, 13, 17, 19, 20, 21 and 22 which are in accordance with the requirements (**Table 2**). Meanwhile, the compounds that meet the requirements for distribution and excretion properties are compounds 1, 2, 4, 7, 8, 9, 11, 12, 13, 17, 18, and 19 (**Table 3**). Therefore, the compounds 9, 11, 13, 17 and 19 can be used as anti-inflammatory drug candidates that have good ADME. Prediction of the metabolism properties of these 22 compounds provides information about the possibility of these

compounds being metabolized in the liver. There are 2 compounds from 5 virtual screening compounds that are predicted to be metabolized in the liver. Compound 11 is a CYP3A4 substrate (M2) and a CYP1A2 inhibitor (M3), while compound 13 is a CYP3A4 substrate (M2) (**Table 4**). If the prediction results of this pharmacokinetic profile are related to previous research on predicting the affinity of compounds in *Hemigraphis alternata* leaves, it can be seen that 3,7,11,15-Tetramethyl-2-hexadecen-1-ol is a compound that has good activity and ADME for the inflammation treatment.

Table 2. The prediction results of absorption properties 22 compounds contained in Hemigraphis alternata using pkCSM

			mata asii	ig proon				
No	Compound	A1	A2	А3	A4	A5	A6	A7
1	15-Chloro-4-pentadecyne	-7.634	1.402	92.577	-2.420	No	No	No
2	4-(2-Methoxyphenyl)piperidine	-1.835	1.385	91.872	-2.283	No	No	No
3	Cyclobutanol	0.092	1.463	98.450	-3.027	Yes	No	No
4	1-Hexadecyne	-7.801	1.382	92.797	-2.225	No	No	No
5	2-Propylmalonic acid	-1.323	0.667	74.589	-2.735	No	No	No
6	n-Hexadecanoic acid	-5.562	1.558	92.004	-2.717	No	No	No
7	2-Hexylacrylonitrile	-3.861	1.357	94.383	-1.278	No	No	No
8	3,7,11-Trimethyl-1,6,10-	-5.176	1.498	91.887	-1.477	No	No	No
	dodecatrien-3-ol							
9	8a-Methylhexahydro-	-2.187	1.605	97.468	-2.814	No	No	No
	1,8(2H,5H)-naphthalenedione							
10	Acrylonitrile β-[3-(2,2	-4.729	1.382	95.941	-1.606	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
11	3,7,11,15-Tetramethyl-2-	-7.554	1.515	90.710	-2.576	No	No	Yes
	hexadecen-1-ol							
12	Z-2-Dodecenol	-4.816	1.474	91.684	-1.529	No	No	No
13	2-Methylenecholestan-3-ol	-5.818	1.208	95.328	-2.733	No	No	Yes
14	L-Alanine	-2.887	0.466	81.091	-2.738	No	No	No
15	levodopa	-2.890	-0.289	47.741	-2.735	Yes	No	No
16	Glycylsarcosine	-2.699	0.545	68.130	-2.735	No	No	No
17	5-Hydroxymethylfurfural	-0.590	1.172	95.848	-3.416	No	No	No
18	10-Undecyn-1-ol	-3.892	1.476	93.273	-1.448	No	No	No
19	2,5-Dimethyl-2,3-dihydro-5H-	-0.757	1.621	97.700	-2.878	No	No	No
	1,4-dioxepine							
20	4-Nitro-5-hydroxy-1,2-	-2.799	0.903	92.210	-2.622	No	No	No
	dimethylindole							

No	Compound	A1	A2	А3	A4	A5	A6	A7
21	9,9-	-1.452	1.237	100	-3.221	No	No	No
	Dimethoxybicyclo[3.3.1]nona-							
	2,4-dione							
22	2,7-Dioxa-	-1.632	1.563	100	-3.097	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-							
	diene							

Table 3. The prediction results of distribution and Excretion properties 22 compounds contained in *Hemigraphis alternata* using pkCSM

No	Compound	D1	D2	D3	D4	E1	E2
1	15-Chloro-4-pentadecyne	0.534	0.062	0.917	-1.257	0.557	No
2	4-(2-Methoxyphenyl)piperidine	1.122	0.462	0.502	-2.260	0.880	No
3	Cyclobutanol	0.047	0.762	-0.031	-2.820	0.448	No
4	1-Hexadecyne	0.631	0.067	0.956	-1.364	1.870	No
5	2-Propylmalonic acid	-0.936	0.588	-0.060	-3.023	0.444	No
6	n-Hexadecanoic acid	-0.543	0.101	-0.111	-1.816	1.763	No
7	2-Hexylacrylonitrile	0.260	0.414	0.571	-1.976	0.550	No
8	3,7,11-Trimethyl-1,6,10-	0.370	0.234	0.652	-2.093	1.739	No
	dodecatrien-3-ol						
9	8a-Methylhexahydro-	0.191	0.564	0.447	-2.813	1.266	No
	1,8(2H,5H)-naphthalenedione						
10	Acrylonitrile β-[3-(2,2	0.531	0.271	0.609	-1.923	0.120	No
	dimethylcyclopropyl)-2,2-						
	dimethylcyclopropyl						
11	3,7,11,15-Tetramethyl-2-	0.468	0	0.806	-1.563	1.686	No
	hexadecen-1-ol						
12	Z-2-Dodecenol	0.358	0.275	0.713	-1.902	1.781	No
13	2-Methylenecholestan-3-ol	-0.145	0	0.808	-1.411	0.546	No
14	L-Alanine	-0.534	0.473	-0.412	-3.405	0.370	No
15	levodopa	-0.105	0.604	-0.843	-3.032	0.430	No
16	Glycylsarcosine	-0.680	0.538	-0.614	-3.183	0.217	No
17	5-Hydroxymethylfurfural	-0.146	0.744	-0.361	-2.914	0.614	No
18	10-Undecyn-1-ol	0.300	0.353	0.721	-1.957	1.713	No
19	2,5-Dimethyl-2,3-dihydro-5H-	-0.007	0.692	0.014	-2.842	0.569	No
	1,4-dioxepine						

No	Compound	D1	D2	D3	D4	E1	E2
20	4-Nitro-5-hydroxy-1,2-	0.209	0.207	-0.263	-2.106	0.537	No
	dimethylindole						
21	9,9-	0.015	0.617	-0.217	-2.909	0.198	No
	Dimethoxybicyclo[3.3.1]nona-						
	2,4-dione						
22	2,7-Dioxa-	0.558	0.678	-0.01	-3.357	0.135	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-						
	diene						

Table 4. The prediction results of metabolism properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	M1	M2	M3	M4	M5	M6	M7
1	15-Chloro-4-pentadecyne	No	Yes	Yes	No	No	No	No
2	4-(2-Methoxyphenyl)piperidine	No	No	No	No	No	No	No
3	Cyclobutanol	No	No	No	No	No	No	No
4	1-Hexadecyne	No	Yes	Yes	No	No	No	No
5	2-Propylmalonic acid	No	No	No	No	No	No	No
6	n-Hexadecanoic acid	No	Yes	No	No	No	No	No
7	2-Hexylacrylonitrile	No	No	No	No	No	No	No
8	3,7,11-Trimethyl-1,6,10-	No	No	No	No	No	No	No
	dodecatrien-3-ol							
9	8a-Methylhexahydro-	No	No	No	No	No	No	No
	1,8(2H,5H)-naphthalenedione							
10	Acrylonitrile β -[3-(2,2	No	No	No	No	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
11	3,7,11,15-Tetramethyl-2-	No	Yes	Yes	No	No	No	No
	hexadecen-1-ol							
12	Z-2-Dodecenol	No	No	No	No	No	No	No
13	2-Methylenecholestan-3-ol	No	Yes	No	No	No	No	No
14	L-Alanine	No	No	No	No	No	No	No
15	levodopa	No	No	No	No	No	No	No
16	Glycylsarcosine	No	No	No	No	No	No	No
17	5-Hydroxymethylfurfural	No	No	No	No	No	No	No
18	10-Undecyn-1-ol	No	No	No	No	No	No	No

No	Compound	M1	M2	M3	M4	M5	M6	M7
19	2,5-Dimethyl-2,3-dihydro-5H-	No	No	No	No	No	No	No
	1,4-dioxepine							
20	4-Nitro-5-hydroxy-1,2-	No	No	Yes	No	No	No	No
	dimethylindole							
21	9,9-	No	No	No	No	No	No	No
	Dimethoxybicyclo[3.3.1]nona-							
	2,4-dione							
22	2,7-Dioxa-	No	No	No	No	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-							
	diene							

CONCLUSION

There are 5 compounds predicted to have the best pharmacokinetic properties in *Hemigraphis* alternata leaves, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine. One of them which is predicted to have high affinity, good anti-inflammatory, absorption, distribution, metabolism and excretion is 3,7,11,15-Tetramethyl-2-hexadecen-1-ol.

ACKNOWLEDGEMENTS

Special thanks to Research and Development Institute of Universitas Muhammadiyah Prof. DR. HAMKA for the support for conducting the research.

CONFLICT OF INTEREST

The authors declare no conflict of interest.

AUTHOR CONTRIBUTIONS

YY and RAR conducted the experiment, YY conducted the conceptualization, methodology, formal analysis, writing-review and editing, RAR conducted data curation and writing-original draft preparation.

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Bukti konfirmasi review dan hasil review pertama (01 Maret 2022)

3/6/25. 12:00 AM #73117 Review

Indonesian Journal of Chem

ISSN: 1411-9420 (print); 2460-1578 (online)

Journal History

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#73117 Review

SUMMARY REVIEW EDITING

Submission

Authors Yeni Yeni, Rizky Arcinthya Rachmania

The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

Leaves Using pkCSM

Section Articles

Editor Stalis Ethica ==

Peer Review

Round 1

Review Version 73117-246939-2-RVDOCX 2022-02-21

Initiated 2022-02-21 2022-03-04 Last modified

Uploaded file Reviewer A 73117-248236-1-RV.DOCX 2022-03-01

Reviewer B 73117-248232-1-RV.DOCX 2022-03-01

Editor Version 73117-247192-1-ED.DOCX 2022-02-21 **Author Version** 73117-248254-1-ED.DOCX 2022-03-01

73117-248254-2-ED.DOCX 2022-03-26 73117-248254-3-ED.DOCX 2022-04-13

Round 2

Review Version 73117-246939-4-RV.DOCX 2022-05-16

2022-04-23 Initiated Last modified 2022-05-07

Uploaded file Reviewer A 73117-254802-1-RV.DOCX 2022-05-06

Editor Decision

Accept Submission 2022-05-16 Decision

Editor/Author Email Record 2022-05-16 Notify Editor

Editor Version 73117-247192-2-ED.DOCX 2022-04-23

73117-247192-3-ED.DOCX 2022-05-07 73117-247192-4-FD.DOCX 2022-05-16

Author Version 73117-248254-4-ED.DOCX 2022-05-13 DELETE

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THE PREDICTION OF PHARMACOKINETIC PROPERTIES OF COMPOUNDS IN HEMIGRAPHIS ALTERNATA (BURM.F.) T. ANDER LEAVES USING PKCSM

ABSTRACT

The inflammatory process has a role in the healing process and leads the body's homeostasis normal. Untreated acute inflammation can lead to organ pathology leading to a chronic inflammatory phenotype. Prediction of the affinity of 22 compounds in *Hemigraphis alternata* leaves as anti-inflammatory has been conducted for COX-1, COX-2 and 5-LOX. The Prediction of pharmacokinetic properties of the compounds was carried out to obtain inflammatory drug candidates that have high affinity with adequate ADME. The application used in this research is pkCSM, a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. The pkCSM used 20 predictors which were divided into 4 properties, absorption (7 predictors), distribution (4 predictors), metabolism (7 predictors) and excretion (2 predictors). Based on the prediction results, there are 5 compounds that have the best pharmacokinetic properties, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine. 3,7,11,15-Tetramethyl-2-hexadecen-1-ol is a compound that is predicted to have high affinity as an anti-inflammatory and good ADME.

Keywords: Pharmacokinetic, Hemigraphis alternata, pkCSM.

INTRODUCTION

Inflammation is a protective reaction of the immune system to defend the body from potentially harmful stimuli, both infectious and non-infectious agents that cause cell damage. It induces the inflammatory cells to be active and trigger inflammatory signaling pathways. The inflammatory process has an important function in the healing process so that abnormal body homeostasis can be restored. Acute inflammation that is not treated properly can cause organ pathology to worsen and eventually lead to a chronic inflammatory phenotype [1–3].

Hemigraphis alternata has various medicinal properties, including anti-inflammatory, anti-nociceptive and anti-diarrhoeal activities. Methanol and ethyl acetate extract of Hemigraphis alternata leaves has been shown to provide anti-inflammatory and non-toxic effects to mice [4]. There are 22 secondary metabolites that have been isolated in the leaves of this plant [5]. These compounds have predicted anti-inflammatory activity against COX-1, COX-2 and 5-LOX receptors. The compound predicted to have the highest affinity for COX-1 and COX-2 were 3,7,11,15-

Tetramethyl-2-hexadecen-1-ol [6]. Meanwhile, the compound predicted to produce the highest affinity for 5-LOX was *n-Hexadecanoic acid* [7].

The relationship of pharmacokinetic properties, toxicity, and potency greatly affect the effectiveness of a drug. Determining the pharmacokinetic profile of a compound is carried out to determine the absorption, distribution, metabolism, and excretion (ADME) properties [8]. The initial assessment of ADME properties will help pharmaceutical researchers to select the best drug candidates for development and reject drug candidates that have a low probability of success. The aim of the in silico prediction of ADME properties is the accurate prediction of the in vivo pharmacokinetic properties of potential drug molecules in humans using only virtual structures [9]. The purpose of this study was to predict the pharmacokinetic properties (ADME) of 22 compounds contained in *Hemigraphis alternata* leaves through an in silico study.

EXPERIMENTAL SECTION

Provide sufficient detail to allow the work to be reproduced, including Materials, Instrumentation, and procedures.

Materials

The SMILES format of 22 compounds contained in the leaves of *Hemigraphis alternata* was obtained from PubChem (pubchem.ncbi.nlm.nih.gov). Compounds that do not have the SMILES format in PubChem can be obtained by describing their structure in the Online SMILES Translator (https://cactus.nci.nih.gov).

Instrumentation

The prediction of pharmacokinetic properties of 22 compounds in the leaves of *Hemigraphis alternata* was conducted using pkCSM (http://biosig.unimelb.edu.au/pkcsm/prediction).

Procedure

pkCSM is based on general properties of compounds (molecular properties, toxicophores and pharmacophores) and distance-based graph signatures. In pkCSM there are 20 predictors that describe the pharmacokinetic properties of a compound. The predictors were divided into absorption of 7 predictors, distribution of 4 predictors, metabolism of 7 predictors and excretion of 2 predictors (**Table 1**) [8,10,11].

Table 1. Distribution of ADME predictors in pkCSM

Property	Predictor	Unit	Property	Predictor	Unit
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Absorption	Water solubility	log mol/L	Distribution	CNS	log PS
	(A1)			permeability	
				(D4)	
Absorption	Caco2	log Papp in	Metabolism	CYP2D6	Yes/No
	permeability (A2)	10-6 cm/s		substrate	
				(M1)	
Absorption	Intestinal	% Absorbed	Metabolism	CYP3A4	Yes/No
	absorption			substrate	
	(human) (A3)			(M2)	
Absorption	Skin Permeability	log Kp	Metabolism	CYP1A2	Yes/No
	(A4)			inhibitor	
				(M3)	
Absorption	P-glycoprotein	Yes/No	Metabolism	CYP2C19	Yes/No
	substrate (A5)			inhibitor	
				(M4)	
Absorption	P-glycoprotein I	Yes/No	Metabolism	CYP2C9	Yes/No
	inhibitor (A6)			inhibitor	
				(M5)	
Absorption	P-glycoprotein II	Yes/No	Metabolism	CYP2D6	Yes/No
	inhibitor (A7)			inhibitor	
				(M6)	
Distribution	VDss (human)	log L/kg	Metabolism	CYP3A4	Yes/No
	(D1)			inhibitor	
				(M7)	
Distribution	Fraction unbound	Fu	Excretion	Total	log
	(human) (D2)			Clearance	ml/min/kg
				(E1)	
Distribution	BBB permeability	log BB	Excretion	Renal	Yes/No
	(D3)	-		OCT2	
				substrate	
				(E2)	

In this study, virtual screening was carried out to obtain several compounds that had good ADME. Virtual screening is based on the results of the ADME predictor which has a numerical value with certain limitations. The predictors were Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3), CNS permeability (D4), total clearance (E1) [12]. The requirements are log Papp in 10-6 cm/s in Caco2 permeability (A2) >

0.9, intestinal absorption (human) (A3) > 30%, log Kp \geq -2.5 in skin permeability (A4), log L/kg \geq -0.15 in VDss (human) (D1), log BB \geq -1 in BBB permeability (D3), log PS \geq -3 in CNS permeability (D4).

The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54, 5 compounds with the best ADME were obtained. Subsequently, these results were compared with the predicted affinity in previous studies and 1 compound with good affinity and ADME was obtained [8].

RESULTS AND DISCUSSION

pkCSM is a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. pkCSM adapts the Cutoff Scanning concept to represent molecular and chemical structures to represent and predict their pharmacokinetic properties [8].

Water solubility is an important factor for a drug to show a better pharmacological response for oral administration. Drugs that have good water solubility properties will cause the drug to have good absorption and bioavailability properties. Good drug absorption and bioavailability can increase the plasma drug concentration at the target site to perform therapeutic functions [13].

Caco-2 cells are a cell line derived from colorectal adenocarcinoma [14]. The Caco-2 model was used to predict the possible gastrointestinal permeability of drugs in pre-clinical trials. This model expresses cytochrome P450 enzymes, transporters, microvilli, and enterocytes based on characteristics identical to those of the human small intestine [15]. The permeability of Caco-2 of a compound is high if it has a Papp $> 8 \times 10$ -6 cm/s. The permeability of Caco-2 is high if the predictive value is > 0.90 in the pkCSM predictive model [8].

The intestine is the main site for absorption of oral drugs. The compound is predicted to have low Intestinal absorption (human) if the value is < 30% [8]. The skin is the boundary between the internal and external body environment. Skin characteristics and properties can modify and affect drug delivery and toxicity [16]. A compound has a tendency to be skin permeable which is expressed by the skin permeability constant logKp (cm/hour). The logKp value of the compound > -2.5 indicates that the compound has low skin permeability [8].

Substrate P-glycoprotein is an ATP-binding cassette (ABC) transporter that functions as a biological barrier by removing toxins and xenobiotics from cells. P-glycoprotein I/II inhibitor is the ability of the compound to inhibit the transport of P-glycoprotein I and P-glycoprotein II. P-glycoprotein-mediated modulation of transport has significant pharmacokinetic implications for Pgp substrates. Inhibition of P-glycoprotein I or P-glycoprotein II can be exploited for certain therapeutic advantages or produce contraindications [8].

VDss (human) is the total amount of drug in the body divided by the total concentration of drug in plasma at steady state. This condition occurs when the system undergoes an infusion of the

drug at a constant rate into the plasma and all drug concentrations in the body do not change [17]. VDss is low if the value is < 0.71 L/kg (log VDss < -0.15) and high if the value is > 2.81 L/kg (log VDss > 0.45). The effectiveness of a drug can be affected by the ability of the drug for binding proteins in the blood. The larger the fraction of the drug that is not bound to protein (fraction unbound), the more efficiently the drug will cross the cell membrane or diffuse [8].

Increased permeability of the BBB (Blood-Brain Barrier) which is a physical and biochemical barrier that plays a role in the defense of cerebral homeostasis can affect the pathological development of ischemic tissue [18,19]. A compound with a logBB > 0.3 can easily pass through BBB and a compound with a logBB < 1 is not well distributed to the brain. The ability of drug compounds to penetrate the CNS (Central Nervous System) can be determined from the value of the blood-brain permeability surface area product (logPS). This value was obtained from in situ brain perfusion with the compound directly injected into the carotid artery without any systemic distribution effect that could distort brain penetration. Compounds with logPS > -2 could penetrate the CNS, while compounds with logPS < -3 could not penetrate the CNS [8].

Cytochrome P450 is a detoxifying enzyme that can be found in the liver. Cytochrome P450 generally plays a role in drug metabolism. However, P450 inhibitors can dramatically alter drug pharmacokinetics. Therefore, it is important to know whether the given compound is a CYP2D6/CYP3A4 substrate predicted to be metabolized by P450. Cytochrome P450 oxidizes xenobiotics to be finally excreted. Many drugs are inactivated by cytochrome P450 and some can be activated by it. These enzyme inhibitors can affect drug metabolism and are contraindicated. It is therefore important to assess the ability of the compound to inhibit cytochrome P450 (isoform CYP1A2/CYP2C19/CYP2D6/CYP3A4). A compound is considered a cytochrome P450 inhibitor if the concentration required to cause 50% inhibition is < 10 M [8].

Drug clearance is the volume of plasma in the vascular compartment that is cleared of drug per unit time. Total clearance is the sum of all body clearances. Total clearance gives an indication of drug elimination from the central compartment without reference to the mechanism of the process [20]. Organic Cation Transporter 2 (OCT2) is a renal uptake transporter that has an important role in the disposition and renal clearance of drugs or endogenous compounds. OCT2 substrates may cause adverse effects when administered concomitantly with OCT2 inhibitors [8].

In the results of the prediction of adsorption properties obtained compounds 3, 6, 9, 11, 13, 17, 19, 20, 21 and 22 which are in accordance with the requirements (**Table 2**). Meanwhile, the compounds that meet the requirements for distribution and excretion properties are compounds 1, 2, 4, 7, 8, 9, 11, 12, 13, 17, 18, and 19 (**Table 3**). Therefore, the compounds 9, 11, 13, 17 and 19 can be used as anti-inflammatory drug candidates that have good ADME. Prediction of the metabolism properties of these 22 compounds provides information about the possibility of these compounds being metabolized in the liver. There are 2 compounds from 5 virtual screening compounds that are predicted to be metabolized in the liver. Compound 11 is a CYP3A4 substrate

(M2) and a CYP1A2 inhibitor (M3), while compound 13 is a CYP3A4 substrate (M2) (**Table 4**). If the prediction results of this pharmacokinetic profile are related to previous research on predicting the affinity of compounds in *Hemigraphis alternata* leaves, it can be seen that 3,7,11,15-Tetramethyl-2-hexadecen-1-ol is a compound that has good activity and ADME for the inflammation treatment.

Table 2. The prediction results of absorption properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	A1	A2	А3	A4	A5	A6	A7
1	15-Chloro-4-pentadecyne	-7.634	1.402	92.577	-2.420	No	No	No
2	4-(2-Methoxyphenyl)piperidine	-1.835	1.385	91.872	-2.283	No	No	No
3	Cyclobutanol	0.092	1.463	98.450	-3.027	Yes	No	No
4	1-Hexadecyne	-7.801	1.382	92.797	-2.225	No	No	No
5	2-Propylmalonic acid	-1.323	0.667	74.589	-2.735	No	No	No
6	n-Hexadecanoic acid	-5.562	1.558	92.004	-2.717	No	No	No
7	2-Hexylacrylonitrile	-3.861	1.357	94.383	-1.278	No	No	No
8	3,7,11-Trimethyl-1,6,10-	-5.176	1.498	91.887	-1.477	No	No	No
	dodecatrien-3-ol							
9	8a-Methylhexahydro-	-2.187	1.605	97.468	-2.814	No	No	No
	1,8(2H,5H)-naphthalenedione							
10	Acrylonitrile β-[3-(2,2	-4.729	1.382	95.941	-1.606	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
11	3,7,11,15-Tetramethyl-2-	-7.554	1.515	90.710	-2.576	No	No	Yes
	hexadecen-1-ol							
12	Z-2-Dodecenol	-4.816	1.474	91.684	-1.529	No	No	No
13	2-Methylenecholestan-3-ol	-5.818	1.208	95.328	-2.733	No	No	Yes
14	L-Alanine	-2.887	0.466	81.091	-2.738	No	No	No
15	levodopa	-2.890	-0.289	47.741	-2.735	Yes	No	No
16	Glycylsarcosine	-2.699	0.545	68.130	-2.735	No	No	No
17	5-Hydroxymethylfurfural	-0.590	1.172	95.848	-3.416	No	No	No
18	10-Undecyn-1-ol	-3.892	1.476	93.273	-1.448	No	No	No
19	2,5-Dimethyl-2,3-dihydro-5H-	-0.757	1.621	97.700	-2.878	No	No	No
	1,4-dioxepine							
20	4-Nitro-5-hydroxy-1,2-	-2.799	0.903	92.210	-2.622	No	No	No
	dimethylindole							

No	Compound	A1	A2	А3	A4	A5	A6	A7
21	9,9-	-1.452	1.237	100	-3.221	No	No	No
	Dimethoxybicyclo[3.3.1]nona-							
	2,4-dione							
22	2,7-Dioxa-	-1.632	1.563	100	-3.097	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-							
	diene							

Table 3. The prediction results of distribution and Excretion properties 22 compounds contained in *Hemigraphis alternata* using pkCSM

No	Compound	D1	D2	D3	D4	E1	E2
1	15-Chloro-4-pentadecyne	0.534	0.062	0.917	-1.257	0.557	No
2	4-(2-Methoxyphenyl)piperidine	1.122	0.462	0.502	-2.260	0.880	No
3	Cyclobutanol	0.047	0.762	-0.031	-2.820	0.448	No
4	1-Hexadecyne	0.631	0.067	0.956	-1.364	1.870	No
5	2-Propylmalonic acid	-0.936	0.588	-0.060	-3.023	0.444	No
6	n-Hexadecanoic acid	-0.543	0.101	-0.111	-1.816	1.763	No
7	2-Hexylacrylonitrile	0.260	0.414	0.571	-1.976	0.550	No
8	3,7,11-Trimethyl-1,6,10-	0.370	0.234	0.652	-2.093	1.739	No
	dodecatrien-3-ol						
9	8a-Methylhexahydro-	0.191	0.564	0.447	-2.813	1.266	No
	1,8(2H,5H)-naphthalenedione						
10	Acrylonitrile β-[3-(2,2	0.531	0.271	0.609	-1.923	0.120	No
	dimethylcyclopropyl)-2,2-						
	dimethylcyclopropyl						
11	3,7,11,15-Tetramethyl-2-	0.468	0	0.806	-1.563	1.686	No
	hexadecen-1-ol						
12	Z-2-Dodecenol	0.358	0.275	0.713	-1.902	1.781	No
13	2-Methylenecholestan-3-ol	-0.145	0	0.808	-1.411	0.546	No
14	L-Alanine	-0.534	0.473	-0.412	-3.405	0.370	No
15	levodopa	-0.105	0.604	-0.843	-3.032	0.430	No
16	Glycylsarcosine	-0.680	0.538	-0.614	-3.183	0.217	No
17	5-Hydroxymethylfurfural	-0.146	0.744	-0.361	-2.914	0.614	No
18	10-Undecyn-1-ol	0.300	0.353	0.721	-1.957	1.713	No
19	2,5-Dimethyl-2,3-dihydro-5H-	-0.007	0.692	0.014	-2.842	0.569	No
	1,4-dioxepine						

No	Compound	D1	D2	D3	D4	E1	E2
20	4-Nitro-5-hydroxy-1,2- dimethylindole	0.209	0.207	-0.263	-2.106	0.537	No
21	9,9- Dimethoxybicyclo[3.3.1]nona- 2,4-dione	0.015	0.617	-0.217	-2.909	0.198	No
22	2,7-Dioxa- tricyclo[4.4.0.0(3,8)]deca-4,9- diene	0.558	0.678	-0.01	-3.357	0.135	No

Table 4. The prediction results of metabolism properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	M1	M2	МЗ	M4	M5	M6	M7
1	15-Chloro-4-pentadecyne	No	Yes	Yes	No	No	No	No
2	4-(2-Methoxyphenyl)piperidine	No	No	No	No	No	No	No
3	Cyclobutanol	No	No	No	No	No	No	No
4	1-Hexadecyne	No	Yes	Yes	No	No	No	No
5	2-Propylmalonic acid	No	No	No	No	No	No	No
6	n-Hexadecanoic acid	No	Yes	No	No	No	No	No
7	2-Hexylacrylonitrile	No	No	No	No	No	No	No
8	3,7,11-Trimethyl-1,6,10-	No	No	No	No	No	No	No
	dodecatrien-3-ol							
9	8a-Methylhexahydro-	No	No	No	No	No	No	No
	1,8(2H,5H)-naphthalenedione							
10	Acrylonitrile β-[3-(2,2	No	No	No	No	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
11	3,7,11,15-Tetramethyl-2-	No	Yes	Yes	No	No	No	No
	hexadecen-1-ol							
12	Z-2-Dodecenol	No	No	No	No	No	No	No
13	2-Methylenecholestan-3-ol	No	Yes	No	No	No	No	No
14	L-Alanine	No	No	No	No	No	No	No
15	levodopa	No	No	No	No	No	No	No
16	Glycylsarcosine	No	No	No	No	No	No	No
17	5-Hydroxymethylfurfural	No	No	No	No	No	No	No
18	10-Undecyn-1-ol	No	No	No	No	No	No	No

No	Compound	M1	M2	М3	M4	M5	M6	M7
19	2,5-Dimethyl-2,3-dihydro-5H-	No	No	No	No	No	No	No
	1,4-dioxepine							
20	4-Nitro-5-hydroxy-1,2-	No	No	Yes	No	No	No	No
	dimethylindole							
21	9,9-	No	No	No	No	No	No	No
	Dimethoxybicyclo[3.3.1]nona-							
	2,4-dione							
22	2,7-Dioxa-	No	No	No	No	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-							
	diene							

CONCLUSION

There are 5 compounds predicted to have the best pharmacokinetic properties in *Hemigraphis alternata* leaves, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine. One of them which is predicted to have high affinity, good anti-inflammatory, absorption, distribution, metabolism and excretion is 3,7,11,15-Tetramethyl-2-hexadecen-1-ol.

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CONFLICT OF INTEREST

The authors declare no conflict of interest.

AUTHOR CONTRIBUTIONS

YY and RAR conducted the experiment, YY conducted the conceptualization, methodology, formal analysis, writing-review and editing, RAR conducted data curation and writing-original draft preparation.

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Review Comments to the article number 73117-246939-2 Indonesian Journal of Chemistry

Reviewer Comments

This article is interesting and written using up to date references but there are some writings that need to be improved according to the specific comments below.

- 1. There are some writing errors, such as capitalization not at the beginning of the sentence (for example in fourth sentence abstract), writing abbreviations that have not been preceded by their abbreviations (for examples COX, LOX, ADME, pkCSM in the abstract).
- 2. This study combines several previous studies. However, in the abstract, it is not clear which one was done by other researchers or what was done by the author of this article. For example, as many as 22 secondary metabolites analyzed for their pharmacokinetic profiles using pkCSM in this article are the result of other people's research. Authors may have conducted previous in silico research to see the predictions of these 22 metabolites as anti-inflammatory (but the in silico docking method used and the results of the in silico research did not explain yet). Furthermore, this study actually only determined pharmacokinetic using pkCSM, but this article also wrote conclusions about compounds that have potential as anti-inflammatory (which were not carried out) and also their good pharmacokinetic parameter. As a suggestion, it is better to make a good common thread regarding the research that has been done previously, the results that have been achieved, and its relationship with the research conducted in this article.
- 3. Try using MESH on demand to search for keywords. For the word pharmacokinetic it is recommended to change it to pharmacokinetic parameters/profiles.
- 4. pkCSM was used as a tool to measure the pharmacokinetic parameters of the secondary metabolite *Hemigraphis alternata*. Therefore, it is necessary to add in the background section about the pkCSM method (their weaknesses/strengths, and add information on the existence of studies showing a positive correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies).
- 5. In addition to the data for compounds whose pharmacokinetic profile will be determined, the structure of secondary metabolites of *Hemigraphis alternata* can be drawn in 3D using Chem bio Draw (accompanied by molecular weight information).
- 6. Table 1 can be made more comprehensive so that it is easier to understand. You can add a column description of the predicted value for each parameter (for example, the

requirements for a good intestinal absorption value > 30/80% and each value can be referenced).

For example:

Pharmacokinetic	Predictor	Unit	Requirement value as	References
Parameter	(Code)		good pharmacokinetic	
			parameter	
Absorption	Water			
	solubility			
Distribution				
	and further			

- 7. Tables 2, 3 and 4 still have to be given a notes/descriptions under each table (information about the predictor code being analyzed) for example D1 = ..., D2 = ... in order to make it easier to read the stand-alone table. In column 2, the name of the compound can be given the molecular weight information.
- 8. The predictor values for each pharmacokinetic parameter have been written in table 1, there is no need to discuss it again in the results and discussion section. It is better to discuss how to select compounds (out of a total of 22 compounds) for each parameters (absorption/distribution/metabolism/excretion) based on the results in table 2-4. For example, a compound is selected if it satisfies all of the criteria for predictor values A1-A7 (absorption parameters), or uses only few important predictor categories.
- 9. This article need to increase the number of references used

3

Bukti konfirmasi submit revisi pertama, respon kepada reviewer, dan artikel yang diresubmit (13 April 2022) 3/6/25. 12:00 AM #73117 Review

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#73117 Review

SUMMARY REVIEW EDITING

Submission

Authors Yeni Yeni, Rizky Arcinthya Rachmania

The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

Leaves Using pkCSM

Section Articles

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Peer Review

Round 1

Review Version 73117-246939-2-RVDOCX 2022-02-21

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Reviewer B 73117-248232-1-RV.DOCX 2022-03-01

Editor Version 73117-247192-1-ED.DOCX 2022-02-21 **Author Version** 73117-248254-1-ED.DOCX 2022-03-01

73117-248254-2-ED.DOCX 2022-03-26 73117-248254-3-ED.DOCX 2022-04-13

Round 2

Review Version 73117-246939-4-RV.DOCX 2022-05-16

2022-04-23 Initiated Last modified 2022-05-07

Uploaded file Reviewer A 73117-254802-1-RV.DOCX 2022-05-06

Editor Decision

Accept Submission 2022-05-16 Decision

Editor/Author Email Record 2022-05-16 Notify Editor

Editor Version 73117-247192-2-ED.DOCX 2022-04-23

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No.	The Reviewer's Comments	Author's Revision
1.	Describe the insilico generally?	The development of new drug
	_ coome and memor goneramy.	candidates is a complex process that
		is expensive and takes a long time.
		Computer-aided drug design (CADD)
		or in silico computational models
		have an important role in the
		discovery of new drugs. Its
		application shortens the research
		time and resources required for the
		rational design of new drug
		candidates. The developments in the
		pharmaceutical field have increased
		the demand for the development of
		more reliable techniques to predict
		the pharmacokinetic properties of new drug candidates. At the present,
		the computational prediction tools
		have been widely used in the drug
		discovery process due to the
		advancement of computational
		algorithms and large knowledge
		databases. In addition, the in silico
		methods have been carried out in
		several drug discoveries currently
		used for the treatment of diseases
2.	Please describe why 5 compound have best	Therefore, the compounds 9 (8a-
	pharmacokinetic in discussion	Methylhexahydro-1,8(2H,5H)-
		naphthalenedione), 11 (3,7,11,15-
		Tetramethyl-2-hexadecen-1-ol), 13
		(2-Methylene cholestan-3-ol), 17 (5-
		Hydroxymethylfurfural) and 19 (2,5-
		Dimethyl-2,3-dihydro-5H-1,4-
		dioxepine) can be used as anti- inflammatory drug candidates that
		have good ADME because these
		compounds are intersection which
		meet the requirements of absorption,
		distribution and excretion predictors.
3.	There are some writing errors, such as	We changed the words COX,
.	capitalization not at the beginning of the	LOX, ADME in abstract. There is
	sentence (for example in fourth sentence	not definite abbreviation for
	abstract), writing abbreviations that have not	pkCSM tool, so that we define it in
	been preceded by their abbreviations (for	the abstract.
	examples COX, LOX, ADME, pkCSM in the	
	abstract).	
4.	This study combines several previous	Hemigraphis alternata is a plant that
4.	studies. However, in the abstract, it is not	
clear which one was done by other		has anti-inflammatory activity. The
	researchers or what was done by the author	compounds contained in Hemigraphis
	of this article. For example, as many as 22	alternata leaves have been predicted
	secondary metabolites analyzed for their	to have affinity for receptors that play
	pharmacokinetic profiles using pkCSM in this	

article are the result of other people's research. Authors may have conducted previous in silico research to see the predictions of these 22 metabolites as antiinflammatory (but the in silico docking method used and the results of the in silico research did not explain yet). Furthermore, this study actually only determined pharmacokinetic using pkCSM, but this article also wrote conclusions about compounds that have potential as antiinflammatory (which were not carried out) and also their good pharmacokinetic parameter. As a suggestion, it is better to make a good common thread regarding the research that has been done previously, the results that have been achieved, and its relationship with the research conducted in this article.

a role in the inflammatory process. A large number of drug candidates were withdrawn from preclinical trials due to their poor pharmacokinetic profiles. Drug compounds must cross the barriers that exist in the body to reach their biological targets so that they can have an effect. The Prediction of pharmacokinetic properties of 22 compounds in Hemigraphis alternata leaves was carried out to obtain inflammatory drug candidates that have adequate pharmacokinetic profiles. The application used in this research is pkCSM, a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distancebased graph signatures. The pkCSM used 20 predictors which were divided into 4 properties, absorption (7 predictors), distribution (4 predictors), metabolism (7 predictors) and excretion (2 predictors). Based on the prediction results, there are 5 compounds that have the best pharmacokinetic properties, 8a-Methylhexahydro-1,8(2H,5H)naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, Methylenecholestan-3-ol, Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4dioxepine.

5. Try using MESH on demand to search for keywords. For the word pharmacokinetic it is

Keywords: *Hemigraphis alternata*, Pharmacokinetic profiles, pkCSM.

	recommended to change it to	
	pharmacokinetic parameters/profiles.	
6.	pkCSM was used as a tool to measure the pharmacokinetic parameters of the secondary metabolite Hemigraphis alternata. Therefore, it is necessary to add in the background section about the pkCSM method (their weaknesses/strengths, and add information on the existence of studies showing a positive correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies).	pkCSM is a tool that can characterize the pharmacokinetic profile of compounds comprehensively. The concept used to predict the predictors by this tool is graph-based structural signatures which train the prediction algorithm by encoding the pattern of distances between atoms. Graphical modeling is the result of an intuitive and well-established mathematical representation of chemical entities. In pkCSM, different predictors including molecular structure and chemistry can be extracted [14–16]. Despite the distribution of experimental values and variability in the size of the data set, pkCSM model was able to achieve a good correlation with experimental results through regression analysis of the ADME
7.	In addition to the data for compounds whose pharmacokinetic profile will be determined, the structure of secondary metabolites of Hemigraphis alternata can be drawn in 3D using Chem bio Draw (accompanied by molecular weight information).	predictors [8]. We do not put the 3D structure of the compounds but 2D structure and molecular weight information because it considers the need to know the atoms that make up the compounds. See Figure 1.
8.	Table 1 can be made more comprehensive so that it is easier to understand. You can add a column description of the predicted value for each parameter (for example, the requirements for a good intestinal absorption value > 30/80% and each value can be	Already fixed. See Table 1.
9.	referenced). Tables 2, 3 and 4 still have to be given a notes/descriptions under each table (information about the predictor code being analyzed) for example D1 =, D2 = in order to make it easier to read the standalone table. In column 2, the name of the compound can be given the molecular weight information.	Already fixed See Table 2, 3 and 4.
10.	The predictor values for each pharmacokinetic parameter have been written in table 1, there is no need to discuss it again in the results and discussion section. It is better to discuss how to select compounds (out of a total of 22 compounds) for each parameters (absorption/distribution/metabolism/excretion)	We have deleted the predictor values in discusion and give more information how to analyze the result of pharmacikinetic properties prediction.

based on the results in table 2-4. For In this study, the screening process example, a compound is selected if it was based on predictors which had a satisfies all of the criteria for predictor values limit value to determine whether or not A1-A7 (absorption parameters), or uses only few important predictor categories. the pharmacokinetic profile of a compound was good. The predictors included Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54. 11 This article need to increase the number of We added 7 references related to references used this research [10] Brogi, S., Ramalho, T.C., Kuca, K., Medina-Franco, J.L. and Valko, M., 2020, In silico methods for drug design and discovery, Front. Chem., 8, 612-616. [11] Chandrasekaran, B., Abed, S.N., Al-Attragchi, O., Kuche, K. and Tekade, R.K., 2018, Computer-aided prediction of pharmacokinetic (ADMET) properties, in dosage form design parameters, pp. 731-755, Academic Press. [12] Shaker, B., Ahmad, S., Lee, J., Jung, C. and Na, D., 2021, In silico methods and tools for drug discovery, Comput. Biol. Med., 137, 104851-104865.

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THE PREDICTION OF PHARMACOKINETIC PROPERTIES OF COMPOUNDS IN HEMIGRAPHIS ALTERNATA (BURM.F.) T. ANDER LEAVES USING PKCSM

Yeni Yeni11,*, and Rizky Arcinthya Rachmania1

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ABSTRACT

The inflammatory process has a role in the healing process and leads the body's homeostasis normal. Untreated acute inflammation can lead to organ pathology leading to a chronic inflammatory phenotype. Hemigraphis alternata is a plant that has anti-inflammatory activity. The compounds contained in Hemigraphis alternata leaves have been predicted to have affinity for receptors that play a role in the inflammatory process. A large number of drug candidates were withdrawn from preclinical trials due to their poor pharmacokinetic profiles. Drug compounds must cross the barriers that exist in the body to reach their biological targets so that they can have an effect. The Prediction of pharmacokinetic properties of 22 compounds in Hemigraphis alternata leaves was carried out to obtain inflammatory drug candidates that have adequate pharmacokinetic profiles. The application used in this research is pkCSM, a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. The pkCSM used 20 predictors which were divided into 4 properties, absorption (7 predictors), distribution (4 predictors), metabolism (7 predictors) and excretion (2 predictors). Based on the prediction results, there are 5 compounds that have the best pharmacokinetic properties, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine.

Keywords: Hemigraphis alternata, Pharmacokinetic profiles, pkCSM.

INTRODUCTION

Inflammation is a protective reaction of the immune system to defend the body from potentially harmful stimuli, both infectious and non-infectious agents that cause cell damage. It induces the inflammatory cells to be active and trigger inflammatory signaling pathways. The inflammatory process has an important function in the healing process so that abnormal body homeostasis can be restored. Acute inflammation that is not treated properly can cause organ pathology to worsen and eventually lead to a chronic inflammatory phenotype [1–3].

Hemigraphis alternata has various medicinal properties, including anti-inflammatory, anti-nociceptive and anti-diarrhoeal activities. Methanol and ethyl acetate extract of Hemigraphis alternata leaves has been shown to provide anti-inflammatory and non-toxic effects to mice [4]. There are 22 secondary metabolites that have been isolated in the leaves of this plant (**Figure 1**) [5]. These compounds have predicted anti-inflammatory activity against COX-1 and 5-LOX receptors [6,7].

The relationship of pharmacokinetic properties, toxicity, and potency greatly affect the effectiveness of a drug. Determining the pharmacokinetic profiles of a compound is carried out to determine the absorption, distribution, metabolism, and excretion (ADME) properties [8]. The initial assessment of ADME properties will help pharmaceutical researchers to select the best drug candidates for development and reject drug candidates that have a low probability of success [9]. The development of new drug candidates is a complex process that is expensive and takes a long time. Computer-aided drug design (CADD) or in silico computational models have an important role in the discovery of new drugs. Its application shortens the research time and resources required for the rational design of new drug candidates. The developments in the pharmaceutical field have increased the demand for the development of more reliable techniques to predict the pharmacokinetic properties of new drug candidates. At the present, the computational prediction tools have been widely used in the drug discovery process due to the advancement of computational algorithms and large knowledge databases. In addition, the in silico methods have been carried out in several drug discoveries currently used for the treatment of diseases [10–13].

pkCSM is a tool that can characterize the pharmacokinetic profile of compounds comprehensively. The concept used to predict the predictors by this tool is graph-based structural signatures which train the prediction algorithm by encoding the pattern of distances between atoms. Graphical modeling is the result of an intuitive and well-established mathematical representation of chemical entities. In pkCSM, different predictors including molecular structure and chemistry can be extracted [14–16]. Despite the distribution of experimental values and variability in the size of the data set, pkCSM model was able to achieve a good correlation with experimental results through regression analysis of the ADME predictors [8]. The aim of the in silico prediction of ADME properties is the accurate prediction of the in vivo pharmacokinetic properties of potential drug molecules in humans using only virtual structures The purpose of this study was to predict the pharmacokinetic properties (ADME) of 22 compounds contained in *Hemigraphis alternata* leaves through an in silico study.

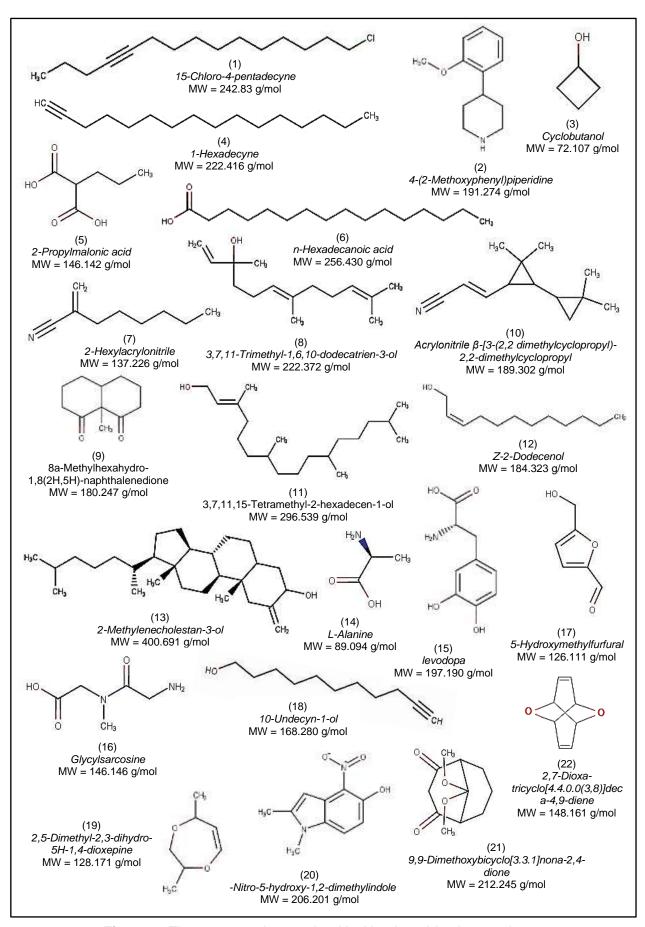


Figure 1. The compounds contained in *Hemigraphis alternata* leaves.

EXPERIMENTAL SECTION

Provide sufficient detail to allow the work to be reproduced, including Materials, Instrumentation, and procedures.

Materials

The SMILES format of 22 compounds contained in the leaves of *Hemigraphis alternata* was obtained from PubChem (pubchem.ncbi.nlm.nih.gov). Compounds that do not have the SMILES format in PubChem can be obtained by describing their structure in the Online SMILES Translator (https://cactus.nci.nih.gov).

Instrumentation

The prediction of pharmacokinetic properties of 22 compounds in the leaves of *Hemigraphis alternata* was conducted using pkCSM (http://biosig.unimelb.edu.au/pkcsm/prediction).

Procedure

pkCSM is based on general properties of compounds (molecular properties, toxicophores and pharmacophores) and distance-based graph signatures. In pkCSM there are 20 predictors that describe the pharmacokinetic properties of a compound. The predictors were divided into absorption of 7 predictors, distribution of 4 predictors, metabolism of 7 predictors and excretion of 2 predictors (**Table 1**) [8,17,18].

Table 1. Distribution of ADME predictors in pkCSM

Pharmacokinetic	Predictor	<u>Unit</u>	Requirement	References
Parameter	(Code)		<mark>value</mark>	
Absorption	Water solubility	log mol/L	-	[8]
	(A1)			
	Caco2 permeability	log Papp in	> 0.9	[8]
	(A2)	10 ⁻⁶ cm/s		
	Intestinal absorption	% Absorbed	> 30%	[8]
	(human)			
	(A3)			
	Skin Permeability	log Kp	≥ -2.5	[8]
	(A4)			
	P-glycoprotein	Yes/No	•	[8]
	<u>substrate</u>			
	(A5)			
	P-glycoprotein I	Yes/No	•	[8]
	<u>inhibitor</u>			
	(A6)			

Pharmacokinetic	Predictor	Unit	Requirement	References
Parameter	(Code)		<u>value</u>	
	P-glycoprotein II	Yes/No	-	[8]
	inhibitor			
	(A7)			
Distribution	VDss (human)	log L/kg	≥ -0.15	<mark>[8]</mark>
	(D1)			
	Fraction unbound	Fu	•	[8]
	(human)			
	(D2)			
	BBB permeability	log BB	<mark>≥ -1</mark>	[8]
	(D3)			
	CNS permeability	log PS	≥ -3	[8]
	(D4)			
Metabolism	CYP2D6 substrate	Yes/No	•	<mark>[8]</mark>
	(M1)		_	
	CYP3A4 substrate	Yes/No	-	<mark>[8]</mark>
	(M2)		_	
	CYP1A2 inhibitor	Yes/No	•	<mark>[8]</mark>
	(M3)		_	
	CYP2C19 inhibitor	Yes/No	•	<mark>[8]</mark>
	(M4)		_	
	CYP2C9 inhibitor	Yes/No	•	<mark>[8]</mark>
	<mark>(M5)</mark>		_	
	CYP2D6 inhibitor	Yes/No	•	[8]
	(M6)		_	
	CYP3A4 inhibitor	Yes/No	•	[8]
	(M7)			-
Excretion	Total Clearance (E1)	log	Higher is better	[8]
		ml/min/kg	_	
	Renal OCT2	Yes/No	•	[8]
	substrate (E2)			

In this study, virtual screening was carried out to obtain several compounds that had good ADME. Virtual screening is based on the results of the ADME predictor which has a numerical value with certain limitations. The predictors were Caco2 permeability (A2), intestinal absorption (human)

(A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3), CNS permeability (D4) and total clearance (E1) [19].

Initially, test compounds were selected based on predictors of Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The compounds that meet the requirements will be re-screened based on the highest total clearance (E1) value [8].

RESULTS AND DISCUSSION

pkCSM is a method for predicting and optimizing the pharmacokinetic properties of small molecules that depend on distance-based graph signatures. pkCSM adapts the Cutoff Scanning concept to represent molecular and chemical structures to represent and predict their pharmacokinetic properties [8].

Water solubility is an important factor for a drug to show a better pharmacological response for oral administration. Drugs that have good water solubility properties will cause the drug to have good absorption and bioavailability properties. Good drug absorption and bioavailability can increase the plasma drug concentration at the target site to perform therapeutic functions [20].

Caco-2 cells are a cell line derived from colorectal adenocarcinoma [21]. The Caco-2 model was used to predict the possible gastrointestinal permeability of drugs in pre-clinical trials. This model expresses cytochrome P450 enzymes, transporters, microvilli, and enterocytes based on characteristics identical to those of the human small intestine [22]. The permeability of Caco-2 of a compound is high if it has a Papp $> 8 \times 10^{-6}$ cm/s.

The intestine is the main site for absorption of oral drugs. The skin is the boundary between the internal and external body environment. Skin characteristics and properties can modify and affect drug delivery and toxicity [23]. A compound has a tendency to be skin permeable which is expressed by the skin permeability constant logKp (cm/hour).

Substrate P-glycoprotein is an ATP-binding cassette (ABC) transporter that functions as a biological barrier by removing toxins and xenobiotics from cells. P-glycoprotein I/II inhibitor is the ability of the compound to inhibit the transport of P-glycoprotein I and P-glycoprotein II. P-glycoprotein-mediated modulation of transport has significant pharmacokinetic implications for Pgp substrates. Inhibition of P-glycoprotein I or P-glycoprotein II can be exploited for certain therapeutic advantages or produce contraindications [8].

VDss (human) is the total amount of drug in the body divided by the total concentration of drug in plasma at steady state. This condition occurs when the system undergoes an infusion of the drug at a constant rate into the plasma and all drug concentrations in the body do not change [24]. The effectiveness of a drug can be affected by the ability of the drug for binding proteins in the blood. The larger the fraction of the drug that is not bound to protein (fraction unbound), the more efficiently the drug will cross the cell membrane or diffuse [8].

Increased permeability of the BBB (Blood-Brain Barrier) which is a physical and biochemical barrier that plays a role in the defense of cerebral homeostasis can affect the pathological development of ischemic tissue [25,26]. The ability of drug compounds to penetrate the CNS (Central Nervous System) can be determined from the value of the blood-brain permeability surface area product (logPS). This value was obtained from in situ brain perfusion with the compound directly injected into the carotid artery without any systemic distribution effect that could distort brain penetration [8].

Cytochrome P450 is a detoxifying enzyme that can be found in the liver. Cytochrome P450 generally plays a role in drug metabolism. However, P450 inhibitors can dramatically alter drug pharmacokinetics. Therefore, it is important to know whether the given compound is a CYP2D6/CYP3A4 substrate predicted to be metabolized by P450. Cytochrome P450 oxidizes xenobiotics to be finally excreted. Many drugs are inactivated by cytochrome P450 and some can be activated by it. These enzyme inhibitors can affect drug metabolism and are contraindicated. It is therefore important to assess the ability of the compound to inhibit cytochrome P450 (isoform CYP1A2/CYP2C19/CYP2C9/CYP2D6/CYP3A4). A compound is considered a cytochrome P450 inhibitor if the concentration required to cause 50% inhibition is < 10 M [8].

Drug clearance is the volume of plasma in the vascular compartment that is cleared of drug per unit time. Total clearance is the sum of all body clearances. Total clearance gives an indication of drug elimination from the central compartment without reference to the mechanism of the process [27]. Organic Cation Transporter 2 (OCT2) is a renal uptake transporter that has an important role in the disposition and renal clearance of drugs or endogenous compounds. OCT2 substrates may cause adverse effects when administered concomitantly with OCT2 inhibitors [8].

In the results of the prediction of absorption properties obtained compounds 3, 6, 9, 11, 13, 17, 19, 20, 21 and 22 which are in accordance with the requirements (**Table 2**). Meanwhile, the compounds that meet the requirements for distribution and excretion properties are compounds 1, 2, 4, 7, 8, 9, 11, 12, 13, 17, 18, and 19 (**Table 3**). Therefore, the compounds 9 (*8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione*), 11 (3,7,11,15-Tetramethyl-2-hexadecen-1-ol), 13 (2-Methylene cholestan-3-ol), 17 (5-Hydroxymethylfurfural) and 19 (2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine) can be used as anti-inflammatory drug candidates that have good ADME because these compounds are intersection which meet the requirements of absorption, distribution and excretion predictors. Prediction of the metabolism properties of these 22 compounds provides information about the possibility of these compounds being metabolized in the liver. There are 2 compounds from 5 virtual screening compounds that are predicted to be metabolized in the liver. Compound 11 is a CYP3A4 substrate (M2) and a CYP1A2 inhibitor (M3), while compound 13 is a CYP3A4 substrate (M2) (**Table 4**). In this study, the screening process was based on predictors which had a limit value to determine whether or not the pharmacokinetic profile of a compound was good. The predictors included Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1),

BBB permeability (D3) and CNS permeability (D4). The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54.

Table 2. The prediction results of absorption properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	MW	A1	A2	A3	A4	A5	<mark>A6</mark>	
1	15-Chloro-4-pentadecyne	242.830	-7.634	1.402	92.577	-2.420	No	No	No
2	4-(2-	191.274	-1.835	1.385	91.872	<mark>-2.283</mark>	No	No	No
	Methoxyphenyl)piperidine								
<mark>3</mark>	Cyclobutanol	<mark>72.107</mark>	0.092	<mark>1.463</mark>	<mark>98.450</mark>	- 3.027	<mark>Yes</mark>	No	No
<mark>4</mark>	1-Hexadecyne	222.416	-7.801	1.382	92.797	<mark>-2.225</mark>	No	No	No
<mark>5</mark>	2-Propylmalonic acid	146.142	- 1.323	0.667	<mark>74.589</mark>	<mark>-2.735</mark>	No	No	No
<mark>6</mark>	n-Hexadecanoic acid	<mark>256.430</mark>	<mark>-5.562</mark>	1.558	92.004	<mark>-2.717</mark>	No	No	No
<mark>7</mark>	2-Hexylacrylonitrile	<mark>137.226</mark>	<mark>-3.861</mark>	<mark>1.357</mark>	94.383	<mark>-1.278</mark>	No	No	No
8	3,7,11-Trimethyl-1,6,10-	<mark>222.372</mark>	<mark>-5.176</mark>	<mark>1.498</mark>	<mark>91.887</mark>	<mark>-1.477</mark>	<mark>No</mark>	No	No
	dodecatrien-3-ol								
9	8a-Methylhexahydro-	180.247	<mark>-2.187</mark>	1.605	<mark>97.468</mark>	<mark>-2.814</mark>	No	<mark>No</mark>	No
	1,8(2H,5H)-								
	<mark>naphthalenedione</mark>								
<mark>10</mark>	Acrylonitrile β -[3-(2,2	<mark>189.302</mark>	<mark>-4.729</mark>	1.382	<mark>95.941</mark>	<mark>-1.606</mark>	No	No	No
	dimethylcyclopropyl)-2,2-								
	dimethylcyclopropyl								
<mark>11</mark>	3,7,11,15-Tetramethyl-2-	<mark>296.539</mark>	<mark>-7.554</mark>	<mark>1.515</mark>	<mark>90.710</mark>	<mark>-2.576</mark>	No	No	Yes
	hexadecen-1-ol								
<mark>12</mark>	Z-2-Dodecenol	184.323	<mark>-4.816</mark>	<mark>1.474</mark>	<mark>91.684</mark>	<mark>-1.529</mark>	No	No	No
<mark>13</mark>	2-Methylenecholestan-3-ol	400.691	<mark>-5.818</mark>	1.208	95.328	<mark>-2.733</mark>	No	No	Yes
<mark>14</mark>	L-Alanine	89.094	<mark>-2.887</mark>	0.466	81.091	<mark>-2.738</mark>	No	No	No
<mark>15</mark>	<i>levodopa</i>	<mark>197.190</mark>	<mark>-2.890</mark>	<mark>-0.289</mark>	47.741	<mark>-2.735</mark>	Yes	No	No
<mark>16</mark>	Glycylsarcosine	<mark>146.146</mark>	<mark>-2.699</mark>	0.545	<mark>68.130</mark>	<mark>-2.735</mark>	No	No	No
<mark>17</mark>	5-Hydroxymethylfurfural	<mark>126.111</mark>	<mark>-0.590</mark>	<mark>1.172</mark>	95.848	<mark>-3.416</mark>	No	No	No
<mark>18</mark>	10-Undecyn-1-ol	168.280	- 3.892	1.476	93.273	<mark>-1.448</mark>	No	No	No
<mark>19</mark>	2,5-Dimethyl-2,3-dihydro-	128.171	<mark>-0.757</mark>	1.621	97.700	<mark>-2.878</mark>	No	No	No
	5H-1,4-dioxepine								
<mark>20</mark>	4-Nitro-5-hydroxy-1,2-	206.201	<mark>-2.799</mark>	0.903	92.210	<mark>-2.622</mark>	No	No	No
	dimethylindole								

No	Compound	MW	A1	A2	A3	A4	A5	A6	A7
<mark>21</mark>	<mark>9,9-</mark>	<mark>212.245</mark>	<mark>-1.452</mark>	1.237	<mark>100</mark>	-3.221	No	No	No
	Dimethoxybicyclo[3.3.1]no								
	<mark>na-2,4-dione</mark>								
<mark>22</mark>	2,7-Dioxa-	148.161	<mark>-1.632</mark>	1.563	<mark>100</mark>	- 3.097	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-								
	4,9-diene								

Note: MW = Molecular Weight (g/mol), A1 = Water solubility, A2 = Caco2 permeability, A3 = Intestinal absorption (human), A4 = Skin Permeability, A5 = P-glycoprotein substrate, A6 = P-glycoprotein I inhibitor, A7 = P-glycoprotein II inhibitor.

Table 3. The prediction results of distribution and Excretion properties 22 compounds contained in *Hemigraphis alternata* using pkCSM

	contained in 776	migrapino	arterriata	using ph	CON		
No	Compound	D1	D2	D3	D4	E1	E2
1	15-Chloro-4-pentadecyne	0.534	0.062	0.917	<mark>-1.257</mark>	<mark>0.557</mark>	No
2	4-(2-Methoxyphenyl)piperidine	1.122	0.462	0.502	<mark>-2.260</mark>	0.880	No
<mark>3</mark>	Cyclobutanol	0.047	<mark>0.762</mark>	-0.031	<mark>-2.820</mark>	<mark>0.448</mark>	No
<mark>4</mark>	1-Hexadecyne	0.631	0.067	0.956	<mark>-1.364</mark>	<mark>1.870</mark>	No
<mark>5</mark>	2-Propylmalonic acid	<mark>-0.936</mark>	<mark>0.588</mark>	<mark>-0.060</mark>	- 3.023	0.444	No
<mark>6</mark>	n-Hexadecanoic acid	<mark>-0.543</mark>	0.101	<mark>-0.111</mark>	<mark>-1.816</mark>	<mark>1.763</mark>	No
<mark>7</mark>	2-Hexylacrylonitrile	0.260	<mark>0.414</mark>	0.571	<mark>-1.976</mark>	<mark>0.550</mark>	No
8	3,7,11-Trimethyl-1,6,10-	<mark>0.370</mark>	<mark>0.234</mark>	<mark>0.652</mark>	<mark>-2.093</mark>	<mark>1.739</mark>	<mark>No</mark>
	dodecatrien-3-ol						
9	8a-Methylhexahydro-	<mark>0.191</mark>	<mark>0.564</mark>	<mark>0.447</mark>	<mark>-2.813</mark>	<mark>1.266</mark>	<mark>No</mark>
	1,8(2H,5H)-naphthalenedione						
<mark>10</mark>	Acrylonitrile β-[3-(2,2	0.531	0.271	<mark>0.609</mark>	<mark>-1.923</mark>	<mark>0.120</mark>	<mark>No</mark>
	dimethylcyclopropyl)-2,2-						
	dimethylcyclopropyl						
<mark>11</mark>	3,7,11,15-Tetramethyl-2-	<mark>0.468</mark>	<mark>0</mark>	<mark>0.806</mark>	<mark>-1.563</mark>	<mark>1.686</mark>	No
	hexadecen-1-ol						
<mark>12</mark>	Z-2-Dodecenol	0.358	<mark>0.275</mark>	0.713	<mark>-1.902</mark>	1.781	No
<mark>13</mark>	2-Methylenecholestan-3-ol	<mark>-0.145</mark>	<mark>0</mark>	<mark>0.808</mark>	<mark>-1.411</mark>	<mark>0.546</mark>	<mark>No</mark>
<mark>14</mark>	L-Alanine	<mark>-0.534</mark>	0.473	<mark>-0.412</mark>	<mark>-3.405</mark>	0.370	No
<mark>15</mark>	<u>levodopa</u>	<mark>-0.105</mark>	0.604	<mark>-0.843</mark>	-3.032	0.430	No
<mark>16</mark>	Glycylsarcosine	<mark>-0.680</mark>	<mark>0.538</mark>	<mark>-0.614</mark>	<mark>-3.183</mark>	0.217	No
<mark>17</mark>	5-Hydroxymethylfurfural	<mark>-0.146</mark>	0.744	-0.361	<mark>-2.914</mark>	<mark>0.614</mark>	No
<mark>18</mark>	10-Undecyn-1-ol	0.300	0.353	0.721	<mark>-1.957</mark>	1.713	No

No	Compound	D1	D2	D3	D4	E1	E2
<mark>19</mark>	2,5-Dimethyl-2,3-dihydro-5H-	-0.007	0.692	0.014	<mark>-2.842</mark>	0.569	No
	1,4-dioxepine						
<mark>20</mark>	4-Nitro-5-hydroxy-1,2-	0.209	0.207	<mark>-0.263</mark>	<mark>-2.106</mark>	<mark>0.537</mark>	No
	<u>dimethylindole</u>						
<mark>21</mark>	<mark>9,9-</mark>	0.015	0.617	- 0.217	<mark>-2.909</mark>	0.198	No
	Dimethoxybicyclo[3.3.1]nona-						
	<mark>2,4-dione</mark>						
<mark>22</mark>	2,7-Dioxa-	0.558	0.678	-0.01	-3.357	0.135	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-						
	<mark>diene</mark>						

Note: D1 = VDss (human), D2 = Fraction unbound (human), D3 = BBB permeability, D4 = CNS permeability, E1 = Total Clearance, E2 = Renal OCT2 substrate.

Table 4. The prediction results of metabolism properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	M1	M2	M3	M4	M5	M6	M7
1	15-Chloro-4-pentadecyne	No	Yes	Yes	No	No	No	No
2	4-(2-Methoxyphenyl)piperidine	No	No	No	No	No	No	No
3	Cyclobutanol	No	No	No	No	No	No	No
4	1-Hexadecyne	No	Yes	Yes	No	No	No	No
<mark>5</mark>	2-Propylmalonic acid	No	No	No	No	No	No	No
<mark>6</mark>	<mark>n-Hexadecanoic acid</mark>	No	Yes	No	No	No	No	No
<mark>7</mark>	2-Hexylacrylonitrile	No	No	No	No	No	No	No
8	3,7,11-Trimethyl-1,6,10-	No	No	No	No	No	No	No
	dodecatrien-3-ol							
9	8a-Methylhexahydro-	No	No	No	No	No	No	No
	1,8(2H,5H)-naphthalenedione							
<mark>10</mark>	Acrylonitrile β-[3-(2,2	No	No	No	No	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
<mark>11</mark>	3,7,11,15-Tetramethyl-2-	No	Yes	Yes	No	No	No	No
	hexadecen-1-ol							
<mark>12</mark>	Z-2-Dodecenol	No	No	No	No	No	No	No
<mark>13</mark>	2-Methylenecholestan-3-ol	No	Yes	No	No	No	No	No
<mark>14</mark>	L-Alanine	No	No	No	No	No	No	No
<mark>15</mark>	<i>levodopa</i>	No	No	No	No	No	No	No

No	Compound	M1	M2	M3	M4	M5	M6	M7
<mark>16</mark>	Glycylsarcosine	No	No	No	No	No	No	No
<mark>17</mark>	5-Hydroxymethylfurfural	No	No	No	No	No	No	No
<mark>18</mark>	10-Undecyn-1-ol	No	No	No	No	No	No	No
<mark>19</mark>	2,5-Dimethyl-2,3-dihydro-5H-	No	No	No	No	No	No	No
	1,4-dioxepine							
<mark>20</mark>	4-Nitro-5-hydroxy-1,2-	No	No	Yes	No	No	No	No
	<u>dimethylindole</u>							
<mark>21</mark>	<mark>9,9-</mark>	No	No	No	No	No	No	No
	Dimethoxybicyclo[3.3.1]nona-							
	<mark>2,4-dione</mark>							
<mark>22</mark>	2,7-Dioxa-	No	No	No	No	No	No	No
	tricyclo[4.4.0.0(3,8)]deca-4,9-							
	<u>diene</u>							

Note: M1 = CYP2D6 substrate, M2 = CYP3A4 substrate, M3 = CYP1A2 inhibitor, M4 = CYP2C19 inhibitor, M5 = CYP2C9 inhibitor, M6 = CYP2D6 inhibitor, M7= CYP3A4 inhibitor.

CONCLUSION

There are 5 compounds predicted to have the best pharmacokinetic properties in *Hemigraphis* alternata leaves, 8a-Methylhexahydro-1,8(2H,5H)-naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1-ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine. These compounds met the most absorption, distribution and excretion predictors requirements compared to other compounds.

ACKNOWLEDGEMENTS

Special thanks to Research and Development Institute of Universitas Muhammadiyah Prof. DR. HAMKA for the support for conducting the research.

CONFLICT OF INTEREST

The authors declare no conflict of interest.

AUTHOR CONTRIBUTIONS

YY and RAR conducted the experiment, YY conducted the conceptualization, methodology, formal analysis, writing-review and editing, RAR conducted data curation and writing-original draft preparation.

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Bukti konfirmasi review dan hasil review kedua (06 Mei 2022)

3/6/25. 12:00 AM #73117 Review

Indonesian Journal of Chem

ISSN: 1411-9420 (print); 2460-1578 (online)

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Authors Yeni Yeni, Rizky Arcinthya Rachmania

The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

Leaves Using pkCSM

Section Articles

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Peer Review

Round 1

Review Version 73117-246939-2-RVDOCX 2022-02-21

Initiated 2022-02-21 2022-03-04 Last modified

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Reviewer B 73117-248232-1-RV.DOCX 2022-03-01

Editor Version 73117-247192-1-ED.DOCX 2022-02-21 **Author Version** 73117-248254-1-ED.DOCX 2022-03-01

73117-248254-2-ED.DOCX 2022-03-26 73117-248254-3-ED.DOCX 2022-04-13

Round 2

Review Version 73117-246939-4-RV.DOCX 2022-05-16

2022-04-23 Initiated Last modified 2022-05-07

Uploaded file Reviewer A 73117-254802-1-RV.DOCX 2022-05-06

Editor Decision

Accept Submission 2022-05-16 Decision

Editor/Author Email Record 2022-05-16 Notify Editor

Editor Version 73117-247192-2-ED.DOCX 2022-04-23

73117-247192-3-ED.DOCX 2022-05-07 73117-247192-4-FD.DOCX 2022-05-16

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No.	The Reviewer's First Comments	Author's Revision	Second Comments
1.	There are some writing errors, such as capitalization not at the beginning of the sentence (for example in fourth sentence abstract), writing abbreviations that have not been preceded by their abbreviations (for examples COX, LOX, ADME, pkCSM in the abstract).	We changed the words COX, LOX, ADME in abstract. There is not definite abbreviation for pkCSM tool, so that we define it in the abstract. Hemigraphis alternata is a plant that	Writing corrections are not done thoroughly, capitalization not at the beginning of the sentences is still visible in the abstract and the whole article. In addition, the abbreviation must be fully written first (examples COX, LOX).
2.	This study combines several previous studies. However, in the abstract, it is not clear which one	has anti-inflammatory activity. The	The author can't answer the first comment clearly.
	was done by other researchers or what was done by the author of this article. For example, as many as 22 secondary metabolites analyzed for	compounds contained in <i>Hemigraphis alternata</i> leaves have been predicted	
	their pharmacokinetic profiles using pkCSM in	to have affinity for receptors that play	
	this article are the result of other people's research. Authors may have conducted previous	a role in the inflammatory process. A	
	in silico research to see the predictions of these 22 metabolites as anti-inflammatory (but the in silico docking method used and the results of the	large number of drug candidates were	
		withdrawn from preclinical trials due to	
	in silico research did not explain yet). Furthermore, this study actually only determined	their poor pharmacokinetic profiles.	
	pharmacokinetic using pkCSM, but this article also wrote conclusions about compounds that	Drug compounds must cross the barriers that exist in the body to reach	
	have potential as anti-inflammatory (which were	their biological targets so that they can	
	not carried out) and also their good pharmacokinetic parameter. As a suggestion, it is	have an effect. The Prediction of	
	better to make a good common thread regarding	pharmacokinetic properties of 22	
	the research that has been done previously, the results that have been achieved, and its	compounds in Hemigraphis alternata	
	relationship with the research conducted in this article.	leaves was carried out to obtain	
		inflammatory drug candidates that	
		have adequate pharmacokinetic	
		profiles. The application used in this	

		research is pkCSM, a method for	
		predicting and optimizing the	
		pharmacokinetic properties of small	
		molecules that depend on distance-	
		based graph signatures. The pkCSM	
		used 20 predictors which were divided	
		into 4 properties, absorption (7	
		predictors), distribution (4 predictors),	
		metabolism (7 predictors) and	
		excretion (2 predictors). Based on the	
		prediction results, there are 5	
		compounds that have the best	
		pharmacokinetic properties, 8a-	
		Methylhexahydro-1,8(2H,5H)-	
		naphthalenedione, 3,7,11,15-	
		Tetramethyl-2-hexadecen-1-ol, 2-	
		Methylenecholestan-3-ol, 5-	
		Hydroxymethylfurfural and 2,5-	
		Dimethyl-2,3-dihydro-5H-1,4-	
		dioxepine.	
3.	Try using MESH on demand to search for	Keywords: Hemigraphis alternata,	
	keywords. For the word pharmacokinetic it is recommended to change it to pharmacokinetic	Pharmacokinetic profiles, pkCSM.	
	parameters/profiles.		

4.	pkCSM was used as a tool to measure the pharmacokinetic parameters of the secondary metabolite Hemigraphis alternata. Therefore, it is necessary to add in the background section about the pkCSM method (their weaknesses/strengths, and add information on the existence of studies showing a positive correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies).	pkCSM is a tool that can characterize the pharmacokinetic profile of compounds comprehensively. The concept used to predict the predictors by this tool is graph-based structural signatures which train the prediction algorithm by encoding the pattern of distances between atoms. Graphical modeling is the result of an intuitive and well-established mathematical representation of chemical entities. In pkCSM, different predictors including molecular structure and chemistry can be extracted [14–16]. Despite the distribution of experimental values and variability in the size of the data set, pkCSM model was able to achieve a good correlation with experimental results through regression analysis of the ADME predictors [8].	No add information on the existence of studies showing a positive correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies.
5.	In addition to the data for compounds whose pharmacokinetic profile will be determined, the structure of secondary metabolites of Hemigraphis alternata can be drawn in 3D using Chem bio Draw (accompanied by molecular weight information).	We do not put the 3D structure of the compounds but 2D structure and molecular weight information because it considers the need to know the atoms that make up the compounds. See Figure 1.	No citation in figure 1.
6.	Table 1 can be made more comprehensive so that it is easier to understand. You can add a column description of the predicted value for each parameter (for example, the requirements for a good intestinal absorption value > 30/80% and each value can be referenced).	Already fixed. See Table 1.	Table 1 only cites 1 article.
7.	Tables 2, 3 and 4 still have to be given a notes/descriptions under each table (information about the predictor code being analyzed) for	Already fixed See Table 2, 3 and 4.	

8.	example D1 =, D2 = in order to make it easier to read the stand-alone table. In column 2, the name of the compound can be given the molecular weight information. The predictor values for each pharmacokinetic parameter have been written in table 1, there is no need to discuss it again in the results and discussion section. It is better to discuss how to select compounds (out of a total of 22 compounds) for each parameters (absorption/distribution/metabolism/excretion) based on the results in table 2-4. For example, a compound is selected if it satisfies all of the criteria for predictor values A1-A7 (absorption parameters), or uses only few important predictor categories.	We have deleted the predictor values in discusion and give more information how to analyze the result of pharmacikinetic properties prediction. In this study, the screening process was based on predictors which had a limit value to determine whether or not the pharmacokinetic profile of a compound was good. The predictors included Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The results of the virtual screening were then sorted based on the highest total clearance	The conclusion of choosing a compound that meets the criteria for a pharmacokinetic profile should still be seen from the table presented (table 2-4), for example, by marking which compounds meet the criteria for predictor value. In addition, five selected compound should be highlighted for discussion.
		based on the highest total clearance value (E1) log ml/min/kg, in this study	
9.	This article need to increase the number of	≥ 0.54. We added 7 references related to this	
	references used	research [10] Brogi, S., Ramalho, T.C., Kuca,	

K., Medina-Franco, J.L. and Valko, M., 2020, In silico methods for drug design and discovery, Front. Chem., 8, 612-616. Chandrasekaran, B., Abed, S.N., Al-Attraqchi, O., Kuche, K. and Tekade, R.K., 2018, Computer-aided prediction of pharmacokinetic (ADMET) properties, in dosage form design parameters, pp. 731-755, Academic Press. [12] Shaker, B., Ahmad, S., Lee, J., Jung, C. and Na, D., 2021, In silico methods and tools for drug discovery, Comput. Biol. Med., 137, 104851-104865. [13] de Souza Neto, L. R., Moreira-Filho, J. T., Neves, B. J., Maidana, R. L. B. R., Guimarães, A. C. R., Furnham, N., Andrade, C. H., and Silva, F. P., 2020, In silico strategies

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2021, Advanced bioinformatics		2021, Advanced bioinformatics

	tools in the pharmacokinetic	
	profiles of natural and synthetic	
	compounds with anti-diabetic	
	activity, Biomolecules, 11 (11),	
	1692–1722.	

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Bukti konfirmasi submit revisi kedua, respon kepada reviewer, dan artikel yang diresubmit (13 Mei 2022) 3/6/25. 12:00 AM #73117 Review

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ISSN: 1411-9420 (print); 2460-1578 (online)

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#73117 Review

SUMMARY REVIEW EDITING

Submission

Authors Yeni Yeni, Rizky Arcinthya Rachmania

The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

Leaves Using pkCSM

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capitalization not at the beginning of the sentence (for example in fourth sentence abstract), writing abbreviations that have not been preceded by their abbreviations (for examples COX, LCX, ADME, pk.CSM, and the abstract). **Hemigraphis alternata is a batteriation of the beginning of the sentences is still inflammatory activity. The compounds contained in Hemigraphis alternata leaves have been predicted to have affining to receptors that play a role in the inflammatory process. A large number of drug candidates were withdrawn from preclinical trials due to their poor pharmacokinetic profiles. Drug compounds must cross the bastract waste of the bastract state and the vision of the process of the pharmacokinetic profiles. Drug compounds must cross the bastract state axist in the body to reach their biological targets as that they can have an effect. The Prediction of charmacokinetic profiles in this research is pkCSM, a method for predicting and continuing the pharmacokinetic properties of try and the pharmacokinetic properties of the pharmacokinetic profiles. The application must be fully writing and behavior must be fully writing and behavior and the pharmacokinetic profiles. Drug compounds must cross the bastract state axist in the body to reach their biological targets as that they can have an effect. The Prediction of pharmacokinetic properties of the pharmacokinetic properties of trials after a papilication not at the behavior and the pharmacokinetic profiles. The application used in this research is pkCSM, a method for predictors, and the pharmacokinetic properties of the pharmacokinetic properties of the pharmacokinetic properties of the pharmacokinetic properties of the pharmacokinetic profiles. The application of the pharmacokinetic properties of the pharmacokinetic properties. Between the pharmacokinetic properties of the pharmacokinetic properties. Between the phar	No.	The Reviewer's First	Author's Revision	Second Comments	Second Author's Revision
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analyzed for their pharmacokinetic profiles using pkCSM in this article are the result of other people's research. Authors may have conducted previous in silico research to see the predictions of these 22 metabolites as anti-inflammatory (but the in silico docking method used and the results of the in silico research did not explain yet). Furthermore, this study actually only determined pharmacokinetic using pkCSM, but this article also wrote conclusions about compounds that have potential as antiinflammatory (which were not carried out) and also their good pharmacokinetic parameter. As a suggestion, it is better to make a good common thread regarding the research that has been done previously, the results that have been achieved, and its relationship with the research conducted in this article.

naphthalenedione, 3,7,11,15-Tetramethyl-2-hexadecen-1ol, 2-Methylenecholestan-3-ol, 5-Hydroxymethylfurfural and 2,5-Dimethyl-2,3-dihydro-5H-1,4-dioxepine.

compounds in Hemigraphis alternata (Burm. F.) T. Ander leaves to cyclooxygenase 1 (COX-1): In silico approach, in 4th International Conference on Sustainable Innovation 2020-Health Science and Nursing (ICoSIHSN 2020) January, pp. 552-555, Atlantis Press.) and 5-LOX (Yeni, Y. Rachmania, R. and Yanuar , M.D., 2021, In silico study of compounds contained in Hemigraphis alternata leaves against 5-LOX for antiinflammatory, Indonesia. J. Pharm. Sci. Technol., 8(1),.34-41.). In the abstract and conclusion, we have removed data regarding the predicted affinity of the 22 compounds so that there is no confusion between the results of this study and the results of previous studies. We have previously informed the results of affinity predictions in the introduction section in this article.

INTRODUCTION

Hemigraphis alternata possesses antinociceptive, antiinflammatory and antidiarrheal effects. In mice, ethyl acetate and methanol extracts of Hemigraphis alternata leaves were found to exhibit anti-inflammatory and non-toxic effects [4]. This plant's leaves contain 22 secondary metabolites (Figure 1) [5]. These substances exhibit anti-inflammatory effect against cyclooxygenase-1 (COX-1) and 5lipoxygenase (5-LOX) receptors [6,7].

3. pkCSM was used as a tool to measure the pharmacokinetic parameters of the secondary metabolite Hemigraphis alternata. Therefore, it is necessary to add in the background section about the pkCSM method (their weaknesses/strengths, and add information on the existence of studies showing a positive

pkCSM is a tool that can characterize the pharmacokinetic profile of compounds comprehensively. The concept used to predict the predictors by this tool is graph-based structural signatures which train the prediction algorithm by encoding the pattern of distances between atoms. Graphical modeling is the result of an intuitive and wellestablished mathematical

No add information on the existence of studies showing a positive correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies.

We informed that there is a good correlation between the ADME predictions of pkCSM and the experimental results in the following paragraphs:

INTRODUCTION

Different predictors, including molecular structure and chemistry

			1	
	correlation of the results of pkCSM with the results of in vivo pharmacokinetic studies).	representation of chemical entities. In pkCSM, different predictors including molecular structure and chemistry can be extracted [14–16]. Despite the distribution of experimental values and variability in the size of the data set, pkCSM model was able to achieve a good correlation with experimental results through regression analysis of the ADME		may be retrieved using pkCSM [14–16]. Despite the diversity in the size of the data set and the distribution of experimental values, the pkCSM model was able to establish a strong correlation with experimental results through regression analysis of the ADME predictors [8].
		predictors [8].		In reference [8], the results of regression analysis for absorption predictors are shown considering cross-validation schemes. Pearson's correlation coefficients and standard error are also shown at the top-left corner. The first graph shows the correlation between experimental and predicted values for Caco2 permeability, while the second graph for water solubility.
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4.	In addition to the data for compounds whose pharmacokinetic profile will be determined, the structure of secondary metabolites of Hemigraphis alternata can be drawn in 3D using Chem bio Draw (accompanied by molecular weight information).	We do not put the 3D structure of the compounds but 2D structure and molecular weight information because it considers the need to know the atoms that make up the compounds. See Figure 1.	No citation in figure 1.	Figure 1. The compounds contained in Hemigraphis alternata leaves [6,7].
5.	Table 1 can be made more comprehensive so that it is easier to understand. You can add a column description of the predicted value for each parameter (for example, the requirements for a good intestinal absorption value > 30/80% and each value can be referenced).	Already fixed. See Table 1.	Table 1 only cites 1 article.	In Table 1, there are predictors and their limitations used in this study based on only 1 recommended reference on the pkCSM website (http://biosig.unimelb.edu.au/pkcsm/theory).
6.	The predictor values for each pharmacokinetic parameter have been	We have deleted the predictor values in discusion and give more information how to	The conclusion of choosing a compound that	Table 2-4 highlights the conclusions on the selection of compounds

written in table 1, there is no need to discuss it again in the results and discussion section. It is better to discuss how to select compounds (out of a total of 22 compounds) for each parameters (absorption/distribution/met abolism/excretion) based on the results in table 2-4. For example, a compound is selected if it satisfies all of the criteria for predictor values A1-A7 (absorption parameters), or uses only few important predictor categories.

analyze the result of pharmacikinetic properties prediction.

In this study, the screening process was based on predictors which had a limit value to determine whether or not the pharmacokinetic profile of a compound was good. The predictors included Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54 .

meets the criteria for a pharmacokinetic profile should still be seen from the table presented (table 2-4), for example, by marking which compounds meet the criteria for predictor value. In addition, five selected compound should be highlighted for discussion.

that match the pharmacokinetic profile requirements. The following five compounds are mentioned in this section:

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RESULTS AND DISCUSSION

Therefore, the compounds 9 (8a-methyl-3,4,4a,5,6,7-hexahydro-2H-naphthalene-1,8dione), 11 ((E)-3,7,11,15tetramethylhexadec-2-en-1-ol), 13 (2methylenecholestan-3-ol), 17 (5-(hydroxymethyl) furan-2-carbaldehyde) and 19 (2,3-dihydro-2,5dimethyl-5H-1,4-dioxepin) can be used as antiinflammatory drug candidates that have good ADME because these compounds are intersection which meet the requirements of absorption, distribution and excretion predictors. Prediction of the metabolism properties of these 22 compounds provides information about the possibility of these compounds being metabolized in the liver. There are 2 compounds from 5 virtual screening compounds that are predicted to be metabolized in the liver. Compound 11 is a CYP3A4 substrate (M2) and a CYP1A2 inhibitor (M3), while compound 13 is a CYP3A4 substrate (M2) (Table 4). The basic structures of the five drugs projected to have favorable pharmacokinetic characteristics differ. However, several of them share the same substituents. Compounds 9, 11, 13, and 19 all contain methyl substituents, Compounds 11, 13, and 17 all contain hydroxyl substituents.

THE PREDICTION OF PHARMACOKINETIC PROPERTIES OF COMPOUNDS IN HEMIGRAPHIS ALTERNATA (BURM.F.) T. ANDER LEAVES USING PKCSM

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ABSTRACT

The inflammatory process aids in healing and maintains the body's balance. Untreated acute inflammation can cause organ disease, which can lead to a chronic inflammatory phenotype. *Hemigraphis alternata* is a plant that has anti-inflammatory activity. The compounds contained in *Hemigraphis alternata* leaves have been predicted to have affinity for receptors involved in the inflammatory process. A large number of drug candidates were withdrawn from preclinical trials due to their poor pharmacokinetic profiles. Drug compounds must cross the barriers that exist in the body to reach their biological targets so that they can have an effect. The pharmacokinetic features of 22 components in *Hemigraphis alternata* leaves were predicted in order to produce inflammatory medication candidates with suitable pharmacokinetic profiles. The pkCSM, a strategy for predicting and optimizing the pharmacokinetic properties of tiny molecules based on distance-based graph signatures was used in this work. The pkCSM employed 20 predictors separated into four groups: absorption, distribution, metabolism, and excretion. Based on the prediction findings, there are five substances with the best pharmacokinetic features, *Ba-methyl-3,4,4a,5,6,7-hexahydro-2H-naphthalene-1,8-dione, (E)-3,7,11,15-tetramethylhexadec-2-en-1-ol, 2-methylenecholestan-3-ol, 5-(hydroxymethyl) furan-2-carbaldehyde* and 2,3-dihydro-2,5-dimethyl-5H-1,4-dioxepin.

Keywords: Hemigraphis alternata, Pharmacokinetic profiles, pkCSM.

INTRODUCTION

Inflammation is the body's defensive reaction to potentially hazardous impulses such as viruses or chemicals that induce cell injury. It triggers inflammatory cells and signaling pathways. The process of inflammation is critical in the recovery process because it allows aberrant bodily homeostasis to be restored. Acute inflammation that is not effectively managed can aggravate organ disease and eventually develop to a chronic inflammatory phenotype [1–3].

Hemigraphis alternata possesses anti-nociceptive, anti-inflammatory and anti-diarrheal effects. In mice, ethyl acetate and methanol extracts of Hemigraphis alternata leaves were found to exhibit anti-inflammatory and non-toxic effects [4]. This plant's leaves contain 22 secondary

metabolites (**Figure 1**) [5]. These substances exhibit anti-inflammatory effect against cyclooxygenase-1 (COX-1) and 5-lipoxygenase (5-LOX) receptors [6,7].

The interaction of pharmacokinetic characteristics, toxicity, and potency has a significant impact on a drug's efficacy. A compound's pharmacokinetic profiles are determined to assess its absorption, distribution, metabolism and excretion (ADME) features [8]. The preliminary evaluation of ADME features will assist pharmaceutical researchers in selecting the best medication candidates for development and rejecting drug candidates with a poor likelihood of success [9]. The creation of novel drug candidates is a difficult, time-consuming and expensive procedure. In the development of novel medications, in silico computational model plays an essential role. Its use reduces the amount of time and resources needed for the rational design of novel medication candidates. Pharmaceutical advancements have raised the necessity for more accurate methodologies to predict the pharmacokinetic features of novel drug candidates. Because of the improvement of computer algorithms and massive information databases, computational prediction tools are increasingly routinely employed in the procedure for drug discovery. Furthermore, in silico technologies have been employed in the discovery of various drugs that are now used in the treatment of disorders [10–13].

The application of pkCSM is a tool that can characterize the pharmacokinetic profile of compounds comprehensively. The concept used to predict the predictors by this tool is graph-based structural signatures which train the prediction algorithm by encoding the pattern of distances between atoms. Graphical modeling is the consequence of an understandable and well-established mathematical description of chemical entities. Different predictors, including molecular structure and chemistry may be retrieved using pkCSM [14–16]. Despite the diversity in the size of the data set and the distribution of experimental values, the pkCSM model was able to establish a strong correlation with experimental results through regression analysis of the ADME predictors [8]. The mission of in silico ADME prediction is to accurately forecast the in vivo pharmacokinetic features of prospective therapeutic compounds in humans using only virtual structures. In silico analysis was used to estimate the pharmacokinetic features of 22 chemicals found in *Hemigraphis alternata* leaves.

EXPERIMENTAL SECTION

Include enough information to allow the work to be repeated, such as materials, instruments, and processes.

Materials

PubChem (pubchem.ncbi.nlm.nih.gov) provided the SMILES format of 22 chemicals found in the leaves of *Hemigraphis alternata*. The SMILES translator, cactus in https://cactus.nci.nih.gov can be used to get compounds that do not have the SMILES format in PubChem.

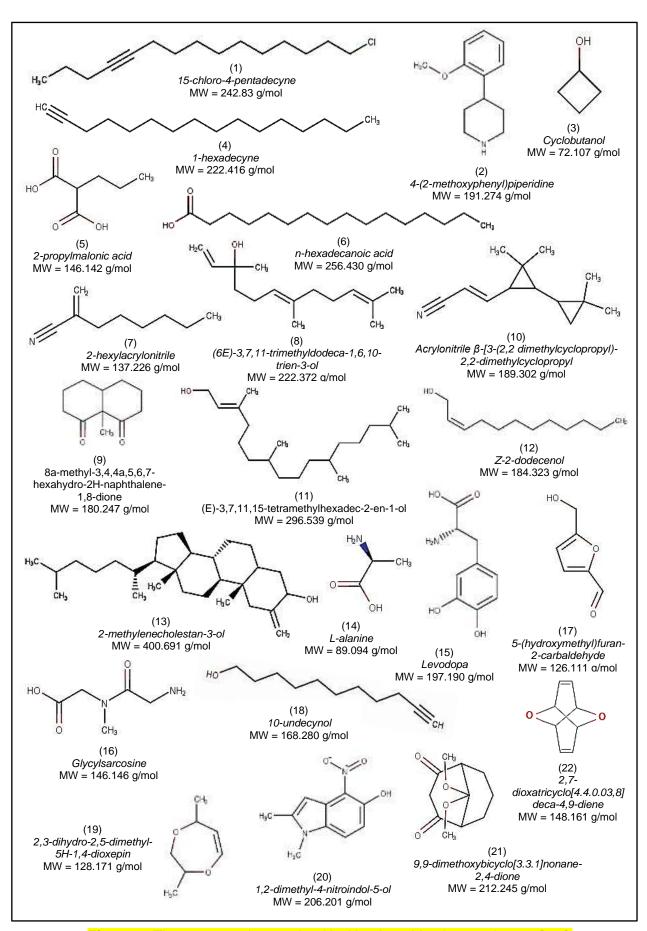


Figure 1. The compounds contained in *Hemigraphis alternata* leaves [6,7].

Instrumentation

The pharmacokinetic characteristics of 22 chemicals in the leaves of *Hemigraphis alternata* were estimated using pkCSM (http://biosig.unimelb.edu.au/pkcsm/prediction).

Procedure

The application of pkCSM is based on compound general qualities (molecular properties, toxicophores, and pharmacophores), as well as distance-based graph signatures. In pkCSM there are 20 predictors that describe the pharmacokinetic properties of a compound. The predictors were divided into absorption of 7 predictors, distribution of 4 predictors, metabolism of 7 predictors and excretion of 2 predictors (**Table 1**) [8,17,18].

Table 1. Distribution of ADME predictors in pkCSM [8]

Pharmacokinetic	Predictor	Unit	Requirement
Parameter	(Code)		value
Absorption	Water solubility	log mol/L	-
	(A1)		
	Caco2 permeability	log Papp in 10 ⁻⁶	> 0.9
	(A2)	cm/s	
	Intestinal absorption (human)	% Absorbed	> 30%
	(A3)		
	Skin permeability	log Kp	≥ -2.5
	(A4)		
	P-glycoprotein substrate	Yes/No	-
	(A5)		
	P-glycoprotein I inhibitor	Yes/No	-
	(A6)		
	P-glycoprotein II inhibitor	Yes/No	-
	(A7)		
Distribution	VDss (human)	log L/kg	≥ -0.15
	(D1)		
	Fraction unbound (human)	Fu	-
	(D2)		
	BBB permeability	log BB	≥ -1
	(D3)		
	CNS permeability	log PS	≥ -3
	(D4)		
Metabolism	CYP2D6 substrate	Yes/No	-
	(M1)		

Pharmacokinetic	Predictor	Unit	Requirement
Parameter	(Code)		value
	CYP3A4 substrate	Yes/No	-
	(M2)		
	CYP1A2 inhibitor	Yes/No	-
	(M3)		
	CYP2C19 inhibitor	Yes/No	-
	(M4)		
	CYP2C9 inhibitor	Yes/No	-
	(M5)		
	CYP2D6 inhibitor	Yes/No	-
	(M6)		
	CYP3A4 inhibitor	Yes/No	-
	(M7)		
Excretion	Total clearance (E1)	log ml/min/kg	Higher is better
	Renal OCT2 substrate (E2)	Yes/No	-

In this study, virtual screening was carried out to obtain several compounds that had good ADME. The findings of the ADME predictor, which has a numerical value with specific constraints, are used in virtual screening. Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), human volume of distribution at steady state (VDss) (D1), Blood-Brain Barrier (BBB) permeability (D3), Central Nervous System (CNS) permeability (D4), and total clearance were the predictors (E1) [19].

Initially, test compounds were selected based on predictors of Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The compounds that meet the requirements will be re-screened based on the highest total clearance (E1) value [8].

RESULTS AND DISCUSSION

The use of pkCSM is a strategy for estimating and improving the pharmacokinetic characteristics of small compounds based on distance-based graph signatures. The use of pkCSM extends the cutoff scanning idea to depict molecular and chemical structures in order to characterize and predict their pharmacokinetic features. [8].

Water solubility is an essential aspect in a drug's pharmacological reaction following oral delivery. Drugs with strong water solubility will have good absorption and bioavailability qualities. Drug absorption and bioavailability can boost plasma drug concentrations at the target location, allowing it to fulfill therapeutic actions [20].

Caco-2 cells are a kind of colorectal cancer cell line [21]. In pre-clinical studies, the Caco-2 model was utilized to predict medication gastrointestinal permeability. Based on the properties of the human small intestine, this model expresses enterocytes, transporters, cytochrome P450 enzymes and microvilli [22]. A compound's Caco-2 permeability is high if its Papp is more than 8 x 10-6 cm/s.

The gut is the primary location of oral medication absorption. The skin serves as a barrier between the interior and exterior environments of the body. Skin qualities and traits can vary and influence medication distribution and toxicity [23]. The skin permeability constant log Kp (cm/hour) expresses a compound's likelihood to be skin permeable.

P-glycoprotein (P-gp) is an ATP-binding cassette (ABC) transporter that acts as a biological barrier in cells by eliminating toxins and xenobiotics. The ability of a chemical to block the transport of P-gp I and P-gp II is referred to as P-gp I/II inhibitor. P-gp-mediated transport modification has important pharmacokinetic consequences for P-gp substrates. Inhibiting P-gp I or P-gp II might provide therapeutic benefits or result in contraindications [8].

VDss (human) is calculated by dividing all drug inside the body by drug concentration in plasma in stable state. This state happens when the system receives a consistent rate of medication infusion into the plasma and all drug concentrations in the body remain constant [24]. The capacity of a medicine to bind proteins in the blood can have an impact on its efficacy. The greater the proportion of the drug that is not bound to protein (fraction unbound), the more effectively the medication will cross or diffuse through the cell membrane [8].

Increased permeability of the BBB, a physical and biochemical barrier that plays a role in the protection of cerebral homeostasis, can alter the pathological development of ischemic tissue [25,26]. The value of the blood-brain permeability surface area product can be used to estimate a pharmacological compound's capacity to reach the CNS (log PS). This value was achieved through in situ brain perfusion with the chemical directly injected into the carotid artery without any systemic distribution impact that may skew brain penetration [8].

Cytochrome P450 is a detoxifying enzyme present in the liver. In general, cytochrome P450 is involved in drug metabolism. However, P450 inhibitors can significantly affect medication pharmacokinetics. As a result, it is critical to determine if the provided molecule is a CYP2D6/CYP3A4 substrate expected to be processed by P450. Cytochrome P450 oxidizes xenobiotics so that they can be excreted. Many medications are inactivated by cytochrome P450, whereas others might be activated by it. These enzyme inhibitors have the potential to interfere with medication metabolism and are thus not recommended. It is therefore critical to evaluate the compound's capacity to inhibit cytochrome P450 (isoforms CYP1A2/CYP2C19/CYP2C9/CYP2D6/CYP3A4). A substance is termed a cytochrome P450 inhibitor if the concentration required to achieve 50% inhibition is less than 10 M. [8].

The amount of drug removed from plasma in the vascular compartment per unit time is referred to as drug clearance. Total clearance is the result of all body clearances. Total clearance indicates drug removal from the core compartment without regard for the process mechanism [27]. The renal uptake transporter Organic Cation Transporter 2 (OCT2) is crucial for disposition of drugs and renal clearance. When used with OCT2 inhibitors, OCT2 substrates might have negative side effects [8].

In the results of the prediction of absorption properties obtained compounds 3, 6, 9, 11, 13, 17, 19, 20, 21 and 22 which are in accordance with the requirements (Table 2). Meanwhile, the compounds that meet the requirements for distribution and excretion properties are compounds 1, 2, 4, 7, 8, 9, 11, 12, 13, 17, 18, and 19 (Table 3). Therefore, the compounds 9 (8a-methyl-3,4,4a,5,6,7-hexahydro-2H-naphthalene-1,8-dione), 11 ((E)-3,7,11,15-tetramethylhexadec-2-en-1ol), 13 (2-methylenecholestan-3-ol), 17 (5-(hydroxymethyl) furan-2-carbaldehyde) and 19 (2,3dihydro-2,5-dimethyl-5H-1,4-dioxepin) can be used as anti-inflammatory drug candidates that have good ADME because these compounds are intersection which meet the requirements of absorption, distribution and excretion predictors. Prediction of the metabolism properties of these 22 compounds provides information about the possibility of these compounds being metabolized in the liver. There are 2 compounds from 5 virtual screening compounds that are predicted to be metabolized in the liver. Compound 11 is a CYP3A4 substrate (M2) and a CYP1A2 inhibitor (M3), while compound 13 is a CYP3A4 substrate (M2) (Table 4). The basic structures of the five drugs projected to have favorable pharmacokinetic characteristics differ. However, several of them share the same substituents. Compounds 9, 11, 13, and 19 all contain methyl substituents. Compounds 11, 13, and 17 all contain hydroxyl substituents. In this study, the screening process was based on predictors which had a limit value to determine whether or not the pharmacokinetic profile of a compound was good. The predictors included Caco2 permeability (A2), intestinal absorption (human) (A3), skin permeability (A4), VDss (human) (D1), BBB permeability (D3) and CNS permeability (D4). The results of the virtual screening were then sorted based on the highest total clearance value (E1) log ml/min/kg, in this study ≥ 0.54 .

Table 2. The prediction results of absorption properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	MW	A1	A2	А3	A4	A5	A6	A7
1	15-chloro-4-pentadecyne	242.830	-7.634	1.402	92.577	-2.420	No	No	No
2	4-(2-	191.274	-1.835	1.385	91.872	-2.283	No	No	No
	methoxyphenyl)piperidine								
3	Cyclobutanol	72.107	0.092	1.463	98.450	-3.027	Yes	No	No
4	1-hexadecyne	222.416	-7.801	1.382	92.797	-2.225	No	No	No
5	2-propylmalonic acid	146.142	-1.323	0.667	74.589	-2.735	No	No	No

6	7
8	0
trimethyldodeca-1,6,10- trien-3-ol 8a-methyl-3,4,4a,5,6,7- hexahydro-2H- naphthalene-1,8-dione 10 Acrylonitrile β-[3-(2,2) dimethylcyclopropyl)-2,2- dimethylcyclopropyl 11 (E)-3,7,11,15- tetramethylhexadec-2-en- 1-ol 12 Z-2-dodecenol 184.323 -4.816 1.474 91.684 -1.529 No No No Yes 14 L-alanine 89.094 -2.887 0.466 81.091 -2.738 No	0
trien-3-ol 9 8a-methyl-3,4,4a,5,6,7-hexahydro-2H-naphthalene-1,8-dione 180.247 -2.187 1.605 97.468 -2.814 No No No 10 Acrylonitrile β-[3-(2,2]] dimethylcyclopropyl)-2,2-dimethylcyclopropyl 189.302 -4.729 1.382 95.941 -1.606 No No No No 11 (E)-3,7,11,15-tetramethylhexadec-2-en-1-ol 296.539 -7.554 1.515 90.710 -2.576 No No No Yes 12 Z-2-dodecenol 184.323 -4.816 1.474 91.684 -1.529 No No No Yes 13 2-methylenecholestan-3-ol 400.691 -5.818 1.208 95.328 -2.733 No No Yes 14 L-alanine 89.094 -2.887 0.466 81.091 -2.738 No No No 15 Levodopa 197.190 -2.890 -0.289 47.741 -2.735 Yes No No 16 Glycylsarcosine 146.146 -2.699 0.545 68.130 -2.735 <t< td=""><td>0</td></t<>	0
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17 <i>5-(hydroxymethyl)furan-2-</i> 126.111 -0.590 1.172 95.848 -3.416 No No No	0
	0
carbaldehvde	0
18 <i>10-undecynol</i> 168.280 -3.892 1.476 93.273 -1.448 No No No	0
19 <i>2,3-dihydro-2,5-dimethyl-</i> 128.171 -0.757 1.621 97.700 -2.878 No No No	0
5H-1,4-dioxepin	
20 1,2-dimethyl-4-nitroindol- 206.201 -2.799 0.903 92.210 -2.622 No No No	0
5-ol	
21 9,9- 212.245 -1.452 1.237 100 -3.221 No No No	0
dimethoxybicyclo[3.3.1]no	
nane-2,4-dione	
22 2,7- 148.161 -1.632 1.563 100 -3.097 No No No	0
dioxatricyclo[4.4.0.03,8]de	
ca-4,9-diene	

Note: = The compounds that satisfy the requirement values, MW = Molecular Weight (g/mol), A1 = Water solubility, A2 = Caco2 permeability, A3 = Intestinal absorption (human), A4 = Skin Permeability, A5 = P-glycoprotein substrate, A6 = P-glycoprotein I inhibitor, A7 = P-glycoprotein II inhibitor.

Table 3. The prediction results of distribution and excretion properties 22 compounds contained in *Hemigraphis alternata* using pkCSM

No	Compound	D1	D2	D3	D4	E1	E2
	Compound						
1	15-chloro-4-pentadecyne	0.534	0.062	0.917	-1.257	0.557	No
2	4-(2-methoxyphenyl)piperidine	1.122	0.462	0.502	-2.260	0.880	No
3	Cyclobutanol	0.047	0.762	-0.031	-2.820	0.448	No
4	1-hexadecyne	0.631	0.067	0.956	-1.364	1.870	No
5	2-propylmalonic acid	-0.936	0.588	-0.060	-3.023	0.444	No
6	n-hexadecanoic acid	-0.543	0.101	-0.111	-1.816	1.763	No
7	2-hexylacrylonitrile	0.260	0.414	0.571	-1.976	0.550	No
8	(6E)-3,7,11-trimethyldodeca-	0.370	0.234	0.652	-2.093	1.739	No
	1,6,10-trien-3-ol						
9	8a-methyl-3,4,4a,5,6,7-	0.191	0.564	0.447	-2.813	1.266	No
	hexahydro-2H-naphthalene-						
	1,8-dione						
10	Acrylonitrile β-[3-(2,2	0.531	0.271	0.609	-1.923	0.120	No
	dimethylcyclopropyl)-2,2-						
	dimethylcyclopropyl						
11	(E)-3,7,11,15-	0.468	0	0.806	-1.563	1.686	No
	tetramethylhexadec-2-en-1-ol						
12	Z-2-dodecenol	0.358	0.275	0.713	-1.902	1.781	No
13	2-methylenecholestan-3-ol	-0.145	0	0.808	-1.411	0.546	No
14	L-alanine	-0.534	0.473	-0.412	-3.405	0.370	No
15	Levodopa	-0.105	0.604	-0.843	-3.032	0.430	No
16	Glycylsarcosine	-0.680	0.538	-0.614	-3.183	0.217	No
17	5-(hydroxymethyl)furan-2-	-0.146	0.744	-0.361	-2.914	0.614	No
	carbaldehyde						
18	10-undecynol	0.300	0.353	0.721	-1.957	1.713	No
19	2,3-dihydro-2,5-dimethyl-5H-	-0.007	0.692	0.014	-2.842	0.569	No
	1,4-dioxepin						
20	1,2-dimethyl-4-nitroindol-5-ol	0.209	0.207	-0.263	-2.106	0.537	No
21	9,9-	0.015	0.617	-0.217	-2.909	0.198	No
	dimethoxybicyclo[3.3.1]nonane-	0.0.0	0.0	J.=	555	000	

dimethoxybicyclo[3.3.1]nonane-

2,4-dione

No	Compound	D1	D2	D3	D4	E1	E2
22	2,7-	0.558	0.678	-0.01	-3.357	0.135	No
	dioxatricyclo[4.4.0.03,8]deca-						
	4,9-diene						

Note: = The compounds that satisfy the requirement values, D1 = VDss (human), D2 = Fraction unbound (human), D3 = BBB permeability, D4 = CNS permeability, E1 = Total Clearance, E2 = Renal OCT2 substrate.

Table 4. The prediction results of metabolism properties 22 compounds contained in Hemigraphis alternata using pkCSM

No	Compound	M1	M2	M3	M4	M5	M6	M7
1	15-chloro-4-pentadecyne	No	Yes	Yes	No	No	No	No
2	4-(2-methoxyphenyl)piperidine	No	No	No	No	No	No	No
3	Cyclobutanol	No	No	No	No	No	No	No
4	1-hexadecyne	No	Yes	Yes	No	No	No	No
5	2-propylmalonic acid	No	No	No	No	No	No	No
6	n-hexadecanoic acid	No	Yes	No	No	No	No	No
7	2-hexylacrylonitrile	No	No	No	No	No	No	No
8	(6E)-3,7,11-trimethyldodeca-	No	No	No	No	No	No	No
	1,6,10-trien-3-ol							
9	8a-methyl-3,4,4a,5,6,7-	No	No	No	No	No	No	No
	hexahydro-2H-naphthalene-							
	1,8-dione							
10	Acrylonitrile β-[3-(2,2	No	No	No	No	No	No	No
	dimethylcyclopropyl)-2,2-							
	dimethylcyclopropyl							
11	(E)-3,7,11,15-	No	Yes	Yes	No	No	No	No
	tetramethylhexadec-2-en-1-ol							
12	Z-2-dodecenol	No	No	No	No	No	No	No
13	2-methylenecholestan-3-ol	No	Yes	No	No	No	No	No
14	L-alanine	No	No	No	No	No	No	No
15	Levodopa	No	No	No	No	No	No	No
16	Glycylsarcosine	No	No	No	No	No	No	No
17	5-(hydroxymethyl)furan-2-	No	No	No	No	No	No	No
	carbaldehyde							
18	10-undecynol	No	No	No	No	No	No	No
19	2,3-dihydro-2,5-dimethyl-5H-	No	No	No	No	No	No	No
	1,4-dioxepin							

No	Compound	M1	M2	M3	M4	M5	M6	M7
20	1,2-dimethyl-4-nitroindol-5-ol	No	No	Yes	No	No	No	No
21	9,9-	No	No	No	No	No	No	No
	dimethoxybicyclo[3.3.1]nonane-							
	2,4-dione							
22	2,7-	No	No	No	No	No	No	No
	dioxatricyclo[4.4.0.03,8]deca-							
	4,9-diene							

Note: ■ = The selected compounds, M1 = CYP2D6 substrate, M2 = CYP3A4 substrate, M3 = CYP1A2 inhibitor, M4 = CYP2C19 inhibitor, M5 = CYP2C9 inhibitor, M6 = CYP2D6 inhibitor, M7= CYP3A4 inhibitor.

CONCLUSION

There are five chemicals in *Hemigraphis alternata* leaves that are projected to have the best pharmacokinetic qualities, *8a-methyl-3,4,4a,5,6,7-hexahydro-2H-naphthalene-1,8-dione, (E)-3,7,11,15-tetramethylhexadec-2-en-1-ol, 2-methylenecholestan-3-ol, 5-(hydroxymethyl) furan-2-carbaldehyde* and *2,3-dihydro-2,5-dimethyl-5H-1,4-dioxepin*. These compounds met the most absorption, distribution and excretion predictors requirements compared to other compounds.

ACKNOWLEDGEMENTS

Universitas Muhammadiyah Prof. DR. HAMKA's Research and Development Institute was especially helpful in performing the research.

CONFLICT OF INTEREST

The authors declare no conflict of interest.

AUTHOR CONTRIBUTIONS

YY and RAR conducted the experiment, YY conducted the conceptualization, methodology, formal analysis, writing-review and editing, RAR conducted data curation and writing-original draft preparation.

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6 Bukti konfirmasi artikel accepted (16 Mei 2022)

3/6/25. 12:00 AM #73117 Review

Indonesian Journal of Chem

ISSN: 1411-9420 (print); 2460-1578 (online)

Journal History

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Authors Yeni Yeni, Rizky Arcinthya Rachmania

The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

Leaves Using pkCSM

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Initiated 2022-02-21 2022-03-04 Last modified

Uploaded file Reviewer A 73117-248236-1-RV.DOCX 2022-03-01

Reviewer B 73117-248232-1-RV.DOCX 2022-03-01

Editor Version 73117-247192-1-ED.DOCX 2022-02-21 **Author Version** 73117-248254-1-ED.DOCX 2022-03-01

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Review Version 73117-246939-4-RV.DOCX 2022-05-16

2022-04-23 Initiated Last modified 2022-05-07

Uploaded file Reviewer A 73117-254802-1-RV.DOCX 2022-05-06

Editor Decision

Accept Submission 2022-05-16 Decision

Editor/Author Email Record 2022-05-16 Notify Editor

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Indonesian Journal of Chem

ISSN: 1411-9420 (print); 2460-1578 (online)

Journal History

Published by Chemistry Department Universitas Gadjah Mada

















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The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander Title

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COPYEDIT INSTRUCTIONS

Copyeditor Dr. Wahyu Dita Saputri

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File: 73117-255666-1-CE.DOCX 2022-05-16

2. **Author Copyedit** 2022-05-31 2022-06-01 2022-06-01

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None

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Layout Comments No Comments

Proofreading

Proofreader Yehezkiel Steven Kurniawan

REVIEW METADATA

REQUEST UNDERWAY COMPLETE Author 2022-07-06 2022-07-06 **2022-07-06** 1. 2022-07-07 2022-07-18 2022-07-18 2 Proofreader 2022-07-18 2022-07-18 2022-07-18 3. Lavout Editor

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Leaves Using pkCSM

Original file 79117-P49999-1-5M/DOC# 2022-02-18

Supp. files None

Submitter Ms Yeni Yeni III

Date submitted February 18, 2022 - 01:32 AM Section Articles Stalls Ethica III

Author Fees

Abstract Views

Article Submission 0.00 USD

Article Publication Paid May 31, 2022 - 08:10 AM

10941

Status

Status Published Vol 22, No 4 (2022)

Initiated 2022-08-10 2022-08-10 Last modified

Submission Metadata

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Google Scholar ID Link https://scholar.google.com/citations?htrenGuser=IEZMHp8AAAAJ

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Country Bio Statement

Title and Abstract

Title The Prediction of Pharmacokinetic Properties of Compounds in Hemigraphis alternata (Burm.F.) T. Ander

Leaves Using pkCSM

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cause organ disease, which can lead to a chronic inflammatory phenotype. Hemigraphis alternata is a plant that has anti-inflammations activity. The community contained in H. alternata leaves have been coefficied to have an

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