

### **ORIGINAL ARTICLE**

# Anti-methanogenic effect of phytogenic extract of *Moringa oleifera* on methane mitigation through inhibition of methyl-coenzyme M reductase receptor: *In silico* study

Muhammad Sulaiman Daulai<sup>1</sup> D, Indah Wijayanti<sup>2</sup> D, Yuli Retnani<sup>2</sup> D, Suzuki Toshisada<sup>3</sup> D

<sup>1</sup>Graduate School of Nutrition and Feed Science, Faculty of Animal Science, IPB University, Bogor, Indonesia

<sup>2</sup>Department of Animal Nutrition and Feed Technology, Faculty of Animal Science, IPB University, Bogor, Indonesia

<sup>3</sup>Department of Biological Molecular Chemistry, Faculty of Agriculture, Kagawa University, Kagawa, Japan

#### **ABSTRACT**

**Objectives:** This study aimed to assess the anti-methanogenic potential of *Moringa oleifera* L. phytogenic extracts through *in silico* inhibition of the methyl-coenzyme M reductase (MCR) receptor. **Materials and Methods:** Phytochemicals from *M. oleifera* were screened and compared with anti-methanogenic compounds such as 3-nitrooxypropanol and native MCR enzyme ligands (coenzyme M and coenzyme B). Molecular docking analysis was performed using AutoDock Vina on PyRx 0.8, and interactions were visualized with Discovery Studio 2024.

**Results:** Selected phytochemicals, including pterygospermin, exhibited promising drug-likeness based on Lipinski's rule of five and absorption, distribution, metabolism, excretion, and toxicology properties. Pterygospermin demonstrated the highest binding affinity to the MCR enzyme's active site, with interactions including Pi-sulfur (Phe443), Pi-alkyl (Val482, Leu320, and Met324), Pi-Pi stacking (Phe330), and van der Waals forces (Tyr333 and Ser325).

**Conclusion:** Pterygospermin shows potential as a competitive inhibitor of the MCR enzyme, providing a sustainable approach to mitigate methane emissions in livestock and contribute to global greenhouse gas reduction efforts.

#### **ARTICLE HISTORY**

Received January 04, 2025 Revised February 15, 2025 Accepted February 17, 2025 Published April 30, 2025

#### **KEYWORDS**

methyl-coenzyme M reductase; Moringa oleifera L.; phytogenic extract



© The authors. This is an Open Access article distributed under the terms of the Creative Commons Attribution 4.0 License (http://creativecommons.org/licenses/by/4.0)

#### Introduction

The livestock industry contributes significantly to greenhouse gas emissions worldwide, primarily through methane (CH<sub>4</sub>) production from enteric fermentation in ruminants, which accounts for approximately 88% of the sector's emissions globally [1]. Methane emissions represent an inefficient utilization of feed energy, as a portion of the feed is converted into CH, and released into the atmosphere, leading to economic losses for farmers and contributing to climate change. With the global demand for animal-derived products projected to double by 2050, driven by increasing living standards, mitigating methane emissions is imperative to enhance livestock productivity and minimize the industry's environmental impact [2]. However, various studies of methane mitigation have been implemented; the anticipated long-term reductions have not been realized. Recently, the majority of approaches to reducing enteric methane emissions have centered on diet and feed additives [3]. In tropical regions, ruminants may benefit significantly from the protein supplements provided by the plant materials, which improve nutritional digestibility and reduce methane emissions. Among these, *Moringa oleifera* stands out due to its widespread availability in tropical and subtropical climates, rapid growth, and high biomass yield. Furthermore, research has demonstrated that dietary supplementation with M. oleifera optimizes microbial metabolic functions and reduces methane emissions. Vitamins, selenium, flavonoids, phenolics, and carotenoids are all abundant in M. oleifera leaves and seed extract, making them a nutritious and healthful substance that can help reduce  $CH_4$  emissions [4].

Identifying key compounds and phytochemicals in M. oleifera with potential anti-methanogenic properties is essential. Many plants have been thoroughly researched

Contact Indah Wijayanti wijayanti@apps.ipb.ac.id Department of Animal Nutrition and Feed Technology, Faculty of Animal Science, IPB University, Bogor, Indonesia.

How to cite this article: Daulai MS, Wijayanti I, Retnani Y, Toshisada S. Anti-methanogenic effect of phytogenic extract of *Moringa oleifera* on methane mitigation through inhibition of methyl-coenzyme M reductase receptor: *In silico* study. J Adv Vet Anim Res 2025; 12(2):361–373.

for their diverse biological activities, and the phytochemicals extracted from *M. oleifera* can be compared with other anti-methanogenic compounds to assess their ability to inhibit critical enzymes involved in methane production in ruminants. It is important to recognize that each plant contains a range of phytochemicals, each with distinct functions and varying impacts on methane emissions. The limited application of computational screening techniques to identify natural inhibitors of methane production has slowed progress, especially since plants like M. oleifera harbor multiple phytochemicals. Advanced computational methods allow for the efficient evaluation of various compounds, overcoming the limitations of labor-intensive in vitro studies and accelerating the discovery process. In silico studies, therefore, present an opportunity to screen numerous phytochemicals against CH, emissions, potentially identifying effective inhibitors without the need for in vitro labor resources [5].

Molecular docking stands as a commonly employed method for *in silico* screening, allowing biomolecules to interact with a target receptor. Therefore, this study focuses on screening compounds to identify prospective inhibitors of the methyl-coenzyme M reductase (MCR) receptor. The MCR enzyme is necessary for methanogenesis in rumen archaea bacteria. As a result, the MCR protein is commonly used in a variety of applications as a methanogenesis marker. Inhibiting MCR can suppress the activity of ruminal methanogens, thereby reducing enteric methane emissions in ruminants [6]. Those phytochemicals showing promise in targeting the enzyme could be candidates for subsequent *in vitro* studies and could be developed as phytogenic feed additives for ruminants. These natural

compounds could reduce methane emissions from livestock, thereby lowering the sector's overall greenhouse gas footprint, contributing to sustainable agricultural practices, and fostering climate resilience. The adoption of such eco-friendly solutions could pave the way for more sustainable systems while reducing reliance on synthetic chemicals, offering both environmental and economic benefits. This study aimed to assess the anti-methanogenic potential of *M. oleifera* L. phytogenic extracts through *in silico* inhibition of the MCR.

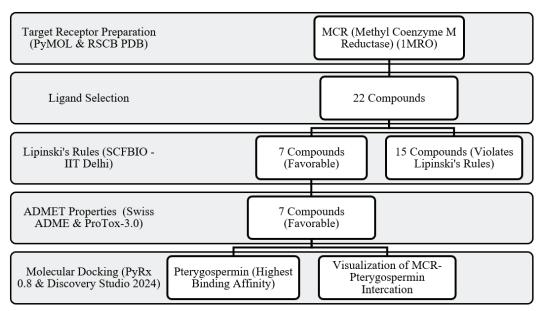
### **Material and Methods**

#### Ligands selection

A total of 22 phytochemicals, which are exclusively derived from the leaves and seeds of *M. oleifera*, were selected from the literature database according to the prior reports by the PubChem database and will be compared with another anti-methanogenic compound [3-nitrooxypropanol (NOP)] and native ligands [coenzyme M (CoM) & coenzyme B (CoB)]. All phytochemical structures were obtained in structured data file and Protein Data Bank (PDB) format from PubChem (pubchem.ncbi.nlm.nih.gov). A workflow of the methodology is presented in Figure 1.

# Lipinski's rule of five

Lipinski's rules were used to assess each ligand's drug-likeness (scfbio-iitd.res.in/software/drugdesign/lipinski.jsp). This rule assesses the possible drug candidate's stability as well as important characteristics such as molecular



**Figure 1.** Workflow diagram of *in silico* study. Adapted with the help of Liu et al. [10].

weight, logP, the number of hydrogen bond donors and acceptors, and molar refractivity [7].

# Absorption, distribution, metabolism, excretion, and toxicology (ADMET) properties analysis

The Swiss ADME tool from the Swiss Institute of Bioinformatics (swissadme.ch) was used to examine the ADMET properties of ligands that satisfy Lipinski's rule of five. The program was used to evaluate the canonical simplified molecular input line entry system, which was acquired from PubChem. Water solubility (log mol/l), lipophilicity (log Po/w), blood-brain barrier (BBB) permeability, gastrointestinal (GI) absorption, and P-gp substrate characteristics were among the attributes that the tool assessed. The Swiss ADME tool utilizes a vector machine algorithm, enabling efficient analysis of datasets for known inhibitors/noninhibitors and substrates/nonsubstrates [8]. Additionally, the evaluation of oral toxicity employed the ProTox 3.0 tool (tox.charite.de/protox3) to predict the lethal dose 50 (LD50) value of compounds. The toxicity results that emerged were then analyzed for acute oral toxicity categories, allowing for a robust preliminary assessment of the toxicological risks associated with oral exposure to the compounds under study [9]. Selected phytochemicals were used for molecular docking analysis.

### Target receptor of study

The RCSB PDB—1MRO (rscb.org/pdb) provided the 3D structure of MCR. PyMOL was used to eliminate any unnecessary water molecules and inhibitors that were attached to the receptor during docking. Energy minimization was performed on both the MCR receptor and the chosen ligands before docking.

# Molecular docking

Phytochemical compounds that did not violate Lipinski's rule of five and exhibited moderate "drug-likeness" were chosen for the molecular docking process. Protein-ligand docking was employed in this study, with MCR as the target receptor. In this study, the docking grid coordinates were carefully chosen to ensure the accuracy and relevance of the in silico molecular docking analysis. The docking grid box was centered at coordinates x = 24.7829, y = 36.2113, and z = -12.9281 to align with the active site of the MCR enzyme. These coordinates correspond to the region where the native ligands, CoM, and CoB, typically bind during the enzymatic process that produces methane in ruminants. The root means square deviation (RMSD) of 0.149 Å was also used as a benchmark for comparing the alignment of docked ligands with the native ligand's conformation, ensuring the reliability of the results. Molecular docking was employed to evaluate the binding affinity and interaction dynamics between phytochemical compounds and the MCR target receptor, which is the crucial enzyme in the rumen that catalyzes methane production, by analyzing the binding energy of the drug-protein complex. AutoDock Vina 1.5.6 from PyRx 0.8 (Virtual Screening Tools) was employed for this analysis, and the results were visualized using the BIOVIA Discovery Studio 2024 tool.

#### **Results**

### Ligands selection

This study revealed that selected ligands in Table 1, classified as selected ligands that were exclusively derived from *M. oleifera*, were evaluated using the PubChem database and relevant literature from Liu et al. [10]. This finding compared selected ligands to comparative ligands (3-NOP) and native ligands, which are substrates of MCR (CoM & CoB).

#### Moringa oleifera compound's drug-likeness properties

The drug-likeness properties of *M. oleifera* were predicted by Lipinski's rule of five. The drug-likeness properties analysis determined the physicochemical characteristics of compounds, including their permeability or ability to diffuse through cell membranes [11]. This result, as shown in Table 2, presents the molar refractivity, log P, number of hydrogen bond donors, number of acceptors, and molecular weight of the selected compounds (Table 2). In line with Lipinski's rule of five, seven molecules were chosen as the best ligands since they satisfied every requirement: 4-[(4'-O-Acetyl-alpha-L-rhamnosyloxy)benzyl]isothiocyanate, niazinin, niazicinin A, pterygospermin, niazirinin, and niazimin.

# ADMET properties of M. oleifera compounds

The ADMET characteristics of seven chosen M. oleifera compounds are revealed in this investigation (Fig. 2). The results demonstrated that Lipinski had neither violated nor approved of the drug-likeness characteristics of 4-[(4'-0-Acetyl-alpha-L-rhamnosyloxy)benzyl] isothiocyanate, Niazinin, Niazicinin A, Pterygospermin, and Niazirinin. All of the substances revealed an appropriate range of lipophilic and hydrophilic characteristics. Pterygospermin and Niazicinin A, on the other hand, demonstrated the highest levels of lipophilicity and water solubility, at 3.75 (Log  $P_{0/w}$ ) and -1.99 (Log mol/l), respectively. Additionally, seven selected compounds showed high GI absorption. The ADME prediction revealed that most of the selected compounds showed negative values of the BBB, which predicted they would not be able to pass through the BBB.

**Table 1.** Selected compound of *M. oleifera*, comparative ligands, and native ligands through virtual screening.

No	Compound	PubChem ID	Sources	
Morin	ga oleifera compound/selected ligand			
1	Quercetin-3'-glucoside	9934142	Seeds, leaves	
2	Niazinin	10088810	Leaves, seeds	
3	Niazimicin A	10247749	Leaves, seeds	
4	O-Methyl-4-[(2',3',4'-tri-O-acetyl-alpha-L-rhamnosyloxy) benzyl]carbamate	101919834	Leaves	
5	O-Ethyl-4-[(2',3',4'-tri-O-acetyl-alpha-L-rhamnosyloxy)benzyl] carbamate	10434741	Leaves	
6	Marumoside A	101794623	Leaves	
7	Pterygospermin	72201063	Seeds, leaves	
8	Moringyne	131751186	Seeds	
9	4-Caffeoylquinic acid;4-O-Caffeoylquinic acid	58427569	Leaves	
10	Niazirinin	10426197	Leaves, seeds	
11	Niazirin	129556	Seeds, leaves	
12	4-[(4'-O-Acetyl-alpha-L-rhamnosyloxy)benzyl]isothiocyanate	10291650	Seeds	
13	Glucomoringin	162639104	Seeds	
14	Glucosinalbin	9601115	Seeds	
15	Glucoraphanin	6602383	Seeds	
16	Glucoiberin	9548622	Seeds	
17	Benzyl glucosinolate	21600402	Seeds	
18	Glucotropaeolin	9548605	Seeds	
19	Glucobarbarin	138756720	Seeds	
20	Glucoraphenin	656559	Seeds	
21	Niazimin	10339912	Leaves, seeds	
22	Niazicinin A	101920262	Leaves, seeds	
Compa	arative ligand			
23	3-NOP	10011893		
Native	ligand			
24	CoM & CoB			

The current investigation shows that niazinin, niazimicin A, and pterygospermin are not P-gp substrates, indicating that their intestinal absorption and bioavailability are likely to be promising [12]. This finding showed that selected compounds of *M. oleifera* are in an acceptable range of LD50 value, which is category III with LD50 values of 500–5,000 mg/kg. In addition, pterygospermin showed the lowest LD50 values of 898 mg/kg, and niazirinin, 4-[(4'-0-acetyl-alpha-L-rhamnosyloxy)benzyl]isothiocyanate, niazimin, and niazicinin A have the highest LD50 values of 4,000 mg/kg. This present study revealed that 7 selected compounds are predicted to be safe for oral administration and do not have lethal effects.

# Molecular docking analysis of M. oleifera compounds

This finding revealed that methyl-CoM and CoB, as the native substrates/ligands of MCR, showed a binding affinity of -5.3 kcal/mol (Fig. 3). In this present study, seven selected compounds of M. oleifera showed higher binding affinity compared to the native ligands (methyl-CoM and CoB), which are niazinin, niazimicin A, pterygospermin, niazirinin, 4-[(4'-O-acetyl-alpha-L-rhamnosyloxy)benzyl] isothiocyanate, niazimin, and niazicinin A, with -5.7, -6.1, -5.8, -7.3, -6.1, -5.6, -5.6, and -5.7 kcal/mol, respectively (Fig. 3). Based on that result, the selected compounds from M. oleifera are promising strong and potential inhibitors for methane production. On the other hand, the comparative

**Table 2.** Selected compounds of *M. oleifera were* analyzed by Lipinski's rule.

No	Compound	Molecular weight (Da)	H acceptor	H donor	Log P	Molar refractivity	
1	Quercetin-3'-glucoside	464.38	12	8	-0.54	110.16	
2	Niazinin	343.40	6	4	0.3	85.68	
3	Niazimicin A	357.42	6	4	0.69	90.49	
4	O-Methyl-4-[(2',3',4'-tri-O-acetyl-alpha- L-rhamnosyloxy)benzyl]carbamate	453.44	10	1	1.85	107.69	
5	O-Ethyl-4-[(2',3',4'-tri-O-acetyl-alpha-L- rhamnosyloxy)benzyl]carbamate	467.47	10	1	2.24	112.50	
6	Marumoside A	297.30	6	4	-0.38	72.11	
7	Pterygospermin	406.52	2	0	4.02	123.32	
8	Moringyne	312.32	7	4	-0.74	75.31	
9	4-Caffeoylquinic acid;4-O-Caffeoylquinic acid	354.31	9	6	-0.65	83.5	
10	Niazirinin	321.33	7	2	0.53	78.68	
11	Niazirin	279.29	6	3	-0.04	68.95	
12	4-[(4'-O-Acetyl-alpha-L-rhamnosyloxy) benzyl]isothiocyanate	353.39	7	2	1.07	88.1	
13	Glucomoringin	609.66	15	7	-2.16	122.7	
14	Glucosinalbin	425.43	11	6	-0.32	93.6	
15	Glucoraphanin	437.51	11	5	0.15	94.59	
16	Glucoiberin	423.48	11	5	-0.24	89.78	
17	Benzyl glucosinolate	408.42	10	4	-0.37	89.72	
18	Glucotropaeolin	409.43	10	5	-0.02	91.57	
19	Glucobarbarin	477.55	11	5	-0.49	95.69	
20	Glucoraphenin	435.49	11	5	0.27	94.12	
21	Niazimin	383.39	8	3	1.1	93.03	
22	Niazicinin A	369.37	8	3	0.71	88.22	
Descrip	otion: = violates Lipinski's rules;	= negative value of Log P;		= favorable compound by Lipinski's rule.			

ligand (3-NOP) showed weak inhibition against MCR with a binding affinity of -3.0 kcal/mol.

Additionally, the Discovery Studio Visualizer, which is shown in Table 3, can be used to visualize the interaction between ligands and macromolecular residues on receptors (MCR). The molecular interactions of pterygospermin with the targeted receptors included several different key residues. This study demonstrates that pterygospermin engages several key residues and a diverse bond network, including Pi-sulfur (Phe443), Pi-alkyl (Val482, Leu320, and Met324), Pi-Pi stacking (Phe330), and van der Waals interactions (Tyr333 and Ser325), all of which support hydrogen bonds and contribute to the stability of the complex [13].

#### **Discussion**

3-NOP has been demonstrated to be a successful feed addition for mitigating enteric methane, and it is advised that ruminants be continually fed it by incorporating it into their daily ration [14]. On the other hand, MCR is the enzyme involved in methane production in methanogenic Archaea. The substrate methyl CoM can enter the enzyme's active sites via a narrow path that is locked once the second substrate, CoB, has bound. In this state, the MCR enzyme contains bound CoM and CoB, and its reactivation is limited to partial recovery through enzymatic reduction [15]. While 3-NOP effectively reduces methane emissions [14], phytochemicals from M. oleifera offer a promising alternative. The comparison between M. oleifera phytochemicals and 3-NOP in methane mitigation can be further expanded by investigating their potential role as methane inhibitors. Certain phytochemicals of M. oleifera may exert

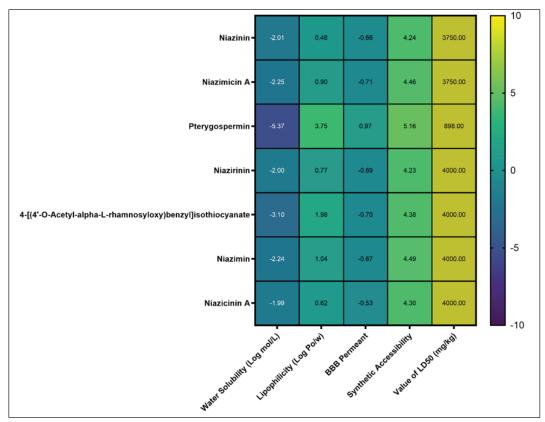


Figure 2. Heatmap displaying the ADMET properties of seven selected compounds from *M. oleifera*.

their effects through a competitive inhibition mechanism against the MCR enzyme. By competing with the natural substrate of MCR, the phytochemicals of *M. oleifera* could interfere with the final step of methanogenesis, thereby reducing methane production. In addition, phytochemicals offer a sustainable alternative to synthetic inhibitors while also providing additional benefits, such as antimicrobial properties, improved animal health, and enhanced feed efficiency [10].

It has been demonstrated that incorporating *M. oleifera* into the diet can control microbial metabolism and reduce methane emissions. In *in vitro* experiments, the inclusion of *M. oleifera* leaf and seed extracts significantly reduced rumen emissions by actively modulating the rumen microbiome. These methane-reducing effects are likely driven by phytochemicals that interact with rumen microbes, altering fermentation patterns and inhibiting methanogen activity [4]. Based on the previous study, the present study employs molecular docking to explore the potential of *M. oleifera* phytochemicals as effective methane inhibitors. This approach to identifying specific compounds capable of targeting MCR provides insights into their mode of action and potential application as natural feed additives for enteric methane reduction.

Drug-likeness is generally characterized by specific physicochemical parameters, including a molecular weight below 500 Daltons (Da), no more than five hydrogen bond donors, fewer than ten hydrogen bond acceptors, a lipophilicity (Log P) value not exceeding five, and a molar refractivity within the range of 40–130 [16]. Drug design focuses significantly on Lipinski's rule of five, and it has been noted that compounds that violate any of these guidelines are likely to have low permeability or poor absorption [17]. The molecules of drugs intended for GI skin penetration should be relatively small, lipophilic, and electrically charge-free. The molar mass must be less than 500 Da in order to penetrate widely through the GI skin [18]. Hydrogen bonds are essential for molecular recognition, drug partitioning, structural stability, enzyme catalysis, and permeability in biochemistry [19]. Additionally, one crucial chemical feature for predicting the oral bioavailability of small drug candidates is the number of hydrogen bond donors and acceptors. Higher hydrogen bond donors or acceptors, however, can have an adverse effect on the drug's permeability and membrane partition [20].

To be effective, a drug needs not only the desired bioactivity but also favorable pharmacokinetics, which requires a careful balance between its Log P and hydrophilicity. Log P, often measured by the Log P value, reflects

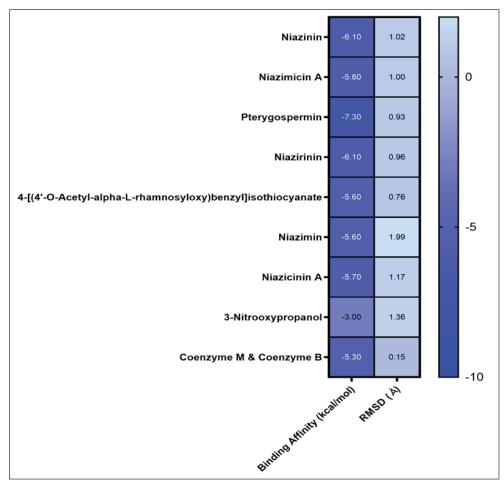


Figure 3. Heatmap displaying the binding affinity of seven selected compounds from *M. oleifera*.

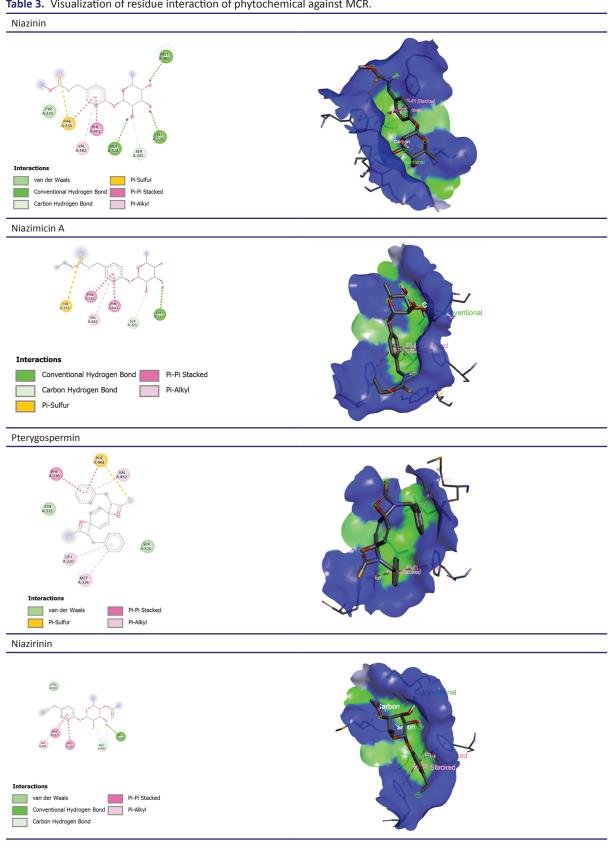
how well a compound distributes between hydrophobic and hydrophilic phases [21]. In addition, the higher Log P value indicates that the molecule is more hydrophobic, making it more likely to be retained in the cell membrane's lipid bilayer. It can lead to a broader distribution of the compound, which might reduce its selectivity for the target [22]. Conversely, the negative value of Log P could not pass through the lipid bilayer membrane, which caused a decrease in compound permeability [23]. Molar refractivity is a steric factor that reveals how small molecules interact spatially within biological environments [24]. Drug solutions' molar refraction and polarizability in water are crucial concepts in pharmaceutical and medicinal chemistry because most biological reactions occur in aqueous environments [25].

Synthesized or isolated compounds were screened against the *in vivo* model to identify potential drugs. However, this method suffered due to the need to screen numerous quantities of molecules, making it labor-intensive. The quantity of active chemicals found was influenced by interactions with the target as well as additional

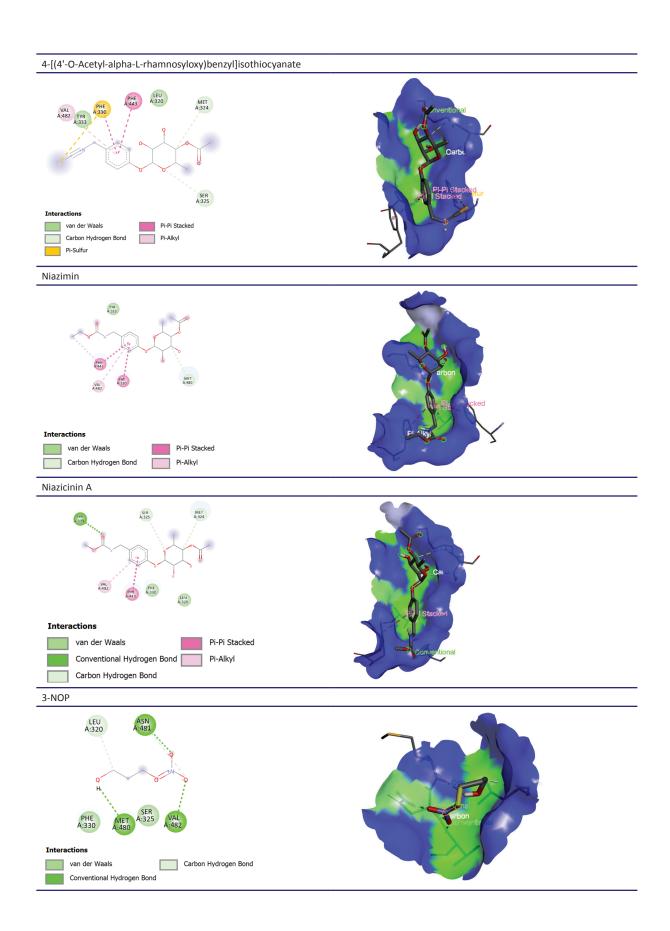
elements like absorption, distribution, and metabolism [26]. Moreover, drug research and discovery require significant investments of energy, resources, and time. Advances in combinatorial chemistry and high-throughput screening have contributed significantly to the number of substances for which early data on ADMET properties are required [27].

Lipophilic molecules are more likely to adopt transcellular transport, while small hydrophobic molecules tend to favor paracellular transport. Other important permeation mechanisms include endocytosis, active transport, and carrier-mediated diffusion in addition to passive diffusion [26]. Based on solubility, Log P, and GI absorption, the ADME characteristics of each molecule indicate that the majority of its constituents have drug-like qualities, with the majority reaching the optimum absorption rate [16]. In addition, seven selected compounds of *M. oleifera*, with a synthetic accessibility score range of 4 and 5 out of 10, fall into the medium range for synthesis, suggesting that it is feasible but may require optimization to reduce production costs. The synthetic process could be streamlined by

Table 3. Visualization of residue interaction of phytochemical against MCR.



(Continued)



identifying and developing efficient synthetic pathways for extracted compounds, enabling its large-scale production. Predicting BBB penetration is essential for determining whether a drug can cross into the brain, which is crucial for reducing side effects, lowering toxicity, or enhancing drug efficacy. While pterygospermin has shown a positive value for BBB penetration, indicating it can cross the barrier, the value is less than 1, implying that it may be inactive in the central nervous system [28].

P-gp serves as a primary obstacle to the effective delivery of drugs, as it actively pumps toxins and foreign substances out of cells [29]. Moreover, the acute toxicity of a compound needs to be assessed to prevent long-term lethal effects. The median LD50, which is the single oral dose anticipated to result in death in 50% of test animals, is the term frequently used to characterize acute oral toxicity [30]. The following are the categories of toxicity: compounds with LD50 values ≤50 mg/kg are included in Category I; those with LD50 values >50 mg/kg but <500 mg/kg are included in Category II; those with LD50 values >500 mg/kg but <5,000 mg/kg are included in Category III; and those with LD50 values >5,000 mg/kg are included in Category IV [31]. Oral LD50 values between 0 and 500 mg/kg (Categories I and II) are regarded as extremely toxic, while LD50 values over 500 mg/kg are classified as having low toxicity [32].

ADMET analysis plays a crucial role in bridging computational predictions with practical implications in livestock research. ADMET screening supports prioritizing drugs with high bioavailability and few side effects by forecasting important pharmacokinetic parameters such as absorption, metabolism, and toxicity. This method enhances the use of resources for in vivo research [33]. This study suggested that seven selected compounds of M. oleifera with high predicted intestinal absorption and low hepatic metabolism are more likely to exhibit desirable pharmacokinetics in livestock, making them strong candidates for further experimental validation. Furthermore, ADMET data refines study designs by informing dosage strategies, metabolic pathways, and safety thresholds, reducing the reliance on trial-and-error approaches. This predictive framework ensures that only the most promising compounds proceed to animal trials, improving research efficiency and cost-effectiveness [34]. Validating ADMET predictions through in vivo studies strengthens the correlation between *in silico* models and biological responses, enhancing the reliability of computational methods for future compound screening.

MCR is the enzyme responsible for methane production in microbes. MCR converts methyl-CoM and CoB into  $\mathrm{CH_4}$  and the heterodisulfide of CoM and CoB. Annually, approximately 109 tons of  $\mathrm{CH_4}$  are produced, which escapes into the atmosphere and contributes significantly as a

greenhouse gas. This makes understanding methane production crucial for environmental studies [15]. The phosphate group of CoB interacts with MCR residues positioned halfway down the channel, specifically with its thiol group from the nickel. The methyl-CoM binding site is located further inside the enzyme, suggesting that this substrate is essential for productive chemical reactions. This has been confirmed by steady-state and single-turnover kinetic studies [35]. It is hypothesized that when both substrates are bound in the active site, CoB induces a conformational rearrangement that positions methyl-CoM in closer proximity to the nickel center, thereby facilitating the cleavage of the carbon-sulfur bond [36].

Predicting the ligand's orientation, location, and conformation within the binding site, as well as determining its binding affinity, are steps in the docking process. Identifying the binding site in advance significantly improves docking efficiency. Docking must determine the most advantageous binding mode within the protein's binding pocket or active site to rank docked ligands appropriately. This requires determining the grid box or central coordinates where ligand-protein interactions occur within the active site [37]. Additionally, [38] revealed that the ligand-receptor interactions are influenced by the binding affinity value, where a lower value of binding affinity indicates a more stable binding between the ligand and receptor. The energy needed for the ligand to connect with the receptor (MCR) binding site is represented by the binding affinity. Several compounds from *M. oleifera* (niazinin, niazimicin A, niazirinin, 4-[(4'-0-Acetyl-alpha-L-rhamnosyloxy)benzyl]isothiocyanate, niazimin, niazicinin) demonstrated favorable ADMET properties, indicating that these compounds are likely to have good oral bioavailability and low toxicity. Furthermore, a ligand location that more closely resembles the natural ligand shape is indicated by a smaller RMSD value. Greater precision in the results is indicated by an RMSD of less than 2 Å, which indicates a reduced calculation error (Fig. 3) [39]. However, not all of them showed high binding affinity compared to pterygospermin (-7.3 kcal/mol). According to this, pterygospermin showed a more effective and stronger inhibition of methane generation, which makes it a viable option for additional research into methane mitigation techniques.

The active site is encircled by Phe330, Tyr333, and Phe443, which are situated close to aromatic and hydrophobic residues [15]. Among all methyl-CoM reductases, these amino acids are conserved [40]. Furthermore, CoB's thiol group interacts with Val482's main chain peptide nitrogen [15]. During the methane generation process, CoB is released, and the activated methyl group absorbs a hydrogen atom. The thiol group of CoB and the peptide nitrogens of Val482 interact in the MCR<sub>ox1-silent</sub> structure to facilitate proton cleavage and the creation of a thiolate

anion. Intermolecular forces such as hydrogen bonds, van der Waals contacts, and electrostatic interactions are crucial for determining the stability and bond energy values of molecular interactions [39]. Methane production inhibitors' biological activity is influenced by these forces, which are especially important when interacting with important residues in the targeted receptor's (MCR) active site (Table 3). This explains why pterygospermin exhibits a higher binding affinity than the native ligands, CoM and CoB, effectively competing with them to inhibit MCR activity. By forming strong intermolecular interactions within the MCR active site, pterygospermin disrupts the final step of methane synthesis in methanogenic archaea, thereby enhancing its inhibitory potential [41].

Pterygospermin, with its strong binding affinity and promising ADMET properties, offers a potential alternative that could outperform existing solutions, particularly in terms of sustainability and cost-effectiveness.

#### Conclusion

The study has identified seven selected compounds derived exclusively from the seeds and leaves of *M. oleifera*, such as niazinin, niazimicin A, pterygospermin, niazirinin, 4-[(4'-0-Acetyl-alpha-L-rhamnosyloxy)benzyl]isothiocyanate, niazimin, and niazicinin A, all of which exhibit promising drug-likeness and favorable ADMET properties. Molecular docking analysis revealed that these seven compounds exhibit higher binding affinities compared to the native ligands, CoM and CoB, indicating strong potential as competitive inhibitors of MCR. Pterygospermin demonstrated the highest binding affinity to the MCR enzyme of -7.3 kcal/mol. Pterygospermin binding interactions with the enzyme's active site include Pi-sulfur (Phe443), Pi-alkyl (Val482, Leu320, and Met324), Pi-Pi stacked (Phe330), and van der Waals interactions (Tyr333 and Ser325). Further experimental validation is needed, particularly through in vitro studies, to confirm the efficacy of pterygospermin as a feed additive for ruminants. Moreover, these findings could facilitate the adoption of plant-based solutions, offering a sustainable, eco-friendly approach to reducing the global greenhouse gas emissions of the livestock industry.

#### List of abbreviations

ADMET, absorption, distribution, metabolism, excretion, and toxicology; BBB, blood-brain barrier; CoB, coenzyme B; CoM, coenzyme M; GI, gastrointestinal; LD50, lethal dose 50; MCR, methyl coenzyme M reductase; NOP, Nitrooxypropanol; PDB, Protein Data Bank; RMSD, root mean square deviation; SDF, Structured data file.

# Acknowledgments

The authors express their sincere gratitude to the Six University Initiative Japan Indonesia (SUIJI) for their invaluable support and the resources made available through the Joint Master Program. They also extend their heartfelt appreciation to the Biological Molecular Chemistry Laboratory, Faculty of Agriculture, Kagawa University, for providing essential facilities, guidance, and a collaborative environment that greatly contributed to this research.

#### **Conflict of interests**

No potential conflict of interest was reported by the author(s).

#### **Authors' contributions**

Study concept and design, acquisition of data, analysis and interpretation of data, drafting of the manuscript, and statistical analysis: MSD. Critical revision of the manuscript for important intellectual content and data collection monitoring: IW, YR, and ST. The final manuscript has been approved for submission by all authors.

#### References

- [1] Thacharodi A, Hassan S, Ahmed TZH, Singh P, Maqbool M, Meenatchi R, et al. The ruminant gut microbiome vs enteric methane emission: the essential microbes may help to mitigate the global methane crisis. Environ Res 2024; 261:119661; https://doi.org/10.1016/j.envres.2024.119661
- [2] Arndt C, Hristov AN, Price WJ, McClelland SC, Pelaez AM, Cueva SF, et al. Full adoption of the most effective strategies to mitigate methane emissions by ruminants can help meet the 1.5 C target by 2030 but not 2050. Proc Natl Acad Sci USA 2022; 119(20):e2111294119; https://doi.org/10.1073/pnas.2111294119
- [3] Honan M, Feng X, Tricario JM, Kebreab E. Feed additives as a strategic approach to reduce enteric methane production in cattle: modes of action, effectiveness and safety. Anim Prod Sci 2022; 62:1303–17; https://doi.org/10.1071/AN20295
- [4] Amad AA, Zentek J. The use of Moringa oleifera in ruminant feeding and its contribution to climate change mitigation. Front Anim Sci 2023; 4:1137562; https://doi.org/10.3389/fanim.2023.1137562
- [5] Dinakarkumar Y, Rajabathar JR, Arokiyaraj S, Jeyaraj I, Anjaneyulu SR, Sandeep S, et al. Anti-methanogenic effect of phytochemicals on methyl-coenzyme m reductase-potential: *in silico* and molecular docking studies for environmental protection. Micromachines 2021; 12(1425):1–14; https://doi.org/10.3390/mi12111425
- [6] Liu Y, Li X, Diao Q, Ma T, Tu Y. In silico and in vitro studies revealed that rosmarinic acid inhibited methanogenesis via regulating composition and function of rumen microbiota. J Dairy Sci 2024; 107:7904–17; https://doi.org/10.3168/jds.2024-24970
- [7] Lipinski CA. Lead- and drug-like compounds: the rule-of-five revolution. Drug Discov Today Technol 2004; 1:337–41; https://doi.org/10.1016/j.ddtec.2004.11.007
- [8] Da Silva CH, Campo VL, Carvalho I, Taft CA. Molecular modeling, docking and ADMET studies applied to the design of a novel

- hybrid for treatment of Alzheimer's disease. J Mol Graph Model 2006; 25:169–75; https://doi.org/10.1016/j.jmgm.2005.12.002
- [9] Cheng F, Li W, Zhou Y, Shen J, Wu Z, Liu G, et al. AdmetSAR: a comprehensive source and free tool for assessment of chemical ADMET properties. J Chem Inf Model 2012; 52(11):3099–105; https://doi.org/10.1021/ci300367a
- [10] Liu R, Liu J, Huang Q, Liu S, Jiang Y. Moringa oleifera: a systematic review of its botany, traditional uses, phytochemistry, pharmacology and toxicity. J Pharm Pharmacol 2021; 74(3):296–320; https://doi.org/10.1093/jpp/rgab131
- [11] Adriani. Prediction of bioactive compounds from Sanrego plants (*Lunasia amara* B.) as inhibitors of the cyclooxygenase-2 enzyme (COX-2) using a molecular docking approach. JIP 2018; 1(1):6–11.
- [12] Snoussi M, Lajimi RH, Badraoui R, Al-Reshidi M, Abdulhakeem MA, Patel M, et al. Chemical composition of *Ducrosia flabellifolia* l. methanolic extract and volatile oil: ADME properties, *in vitro* and *in silico* screening of antimicrobial, antioxidant and anticancer activities. Metabolites 2023; 13(64):1–18; https://doi.org/10.3390/metabo13010064
- [13] Jedli O, Ben-Nasr H, Zammel N, Rebai T, Saoudi M, Elkahoui S, et al. Attenuation of ovalbumin-induced inflammation and lung oxidative injury in asthmatic rats by *Zingiber officinale* extract: combined *in silico* and *in vivo* study on antioxidant potential, STAT6 and TNF-pathways. Biotech 2022; 12:191; https://doi.org/10.1007/s13205-022-03249-5
- [14] Jayanegara A, Sarwono KI, Kondo M, Matsui H, Ridla M, Laconi EB, et al. Use of 3-nitrooxypropanol as feed additive for mitigating enteric methane emissions from ruminants: a meta-analysis. Ital J Anim Sci 2018; 17(3):650-6; https://doi.org/10.1080/18280 51X.2017.1404945
- [15] Ermler U, Grabarse W, Shima S, Goubeaud M, Thauer RK. Crystal structure of methyl-coenzyme m reductase: the key enzyme of biological methane formation. Science 1997; 278:1457–62; https://doi.org/10.1126/science.278.5342.1457
- [16] Diana FAN, Riyadi PH, Susanto E. In silico analysis of ethyl acetate Bruguiera gymnorhiza leaf extracts as an anti-inflammatory agent. J Pengolah Has Perikan Indones 2024; 27(9):798–818; http://dx.doi.org/10.17844/jphpi.v27i9.54255
- [17] Asoka SF, Batubara I, Lestari AR, Wahyuni WT, Wahyudi ST. Compounds in Indonesia ginger rhizome extracts and their potential for anti-skin aging based on molecular docking. Cosmetics 2022; 9(128):1–15; https://doi.org/10.3390/cosmetics9060128
- [18] Souto EB, Fangueiro JF, Fernandes AR, Cano A, Sanchez-Lopez E, Garcia ML, et al. Physicochemical and biopharmaceutical aspects influencing skin permeation and role of SLN and NLC for skin drug delivery. Heliyon 2022; 8(1):1–16; https://doi.org/10.1016/j. heliyon.2022.e08938
- [19] Coimbra JTS, Feghali R, Ribeiro RP, Ramos MJ, Fernandes PA. The importance of intra molecular hydrogen bonds on the translocation of the small drug piracetam through a lipid bilayer. RSC Adv 2021; 11:899–908; https://doi.org/10.1039/D0RA09995C
- [20] Oguma T, Uehara S, Nakahara K, Okuyama Y, Fuchino K, Suzuki N, et al. A quantum mechanics-based method to predict intra-molecular hydrogen bond formation reflecting p-glycoprotein recognition. ACS Med Chem Lett 2023; 14(2):223–8; https://doi.org/10.1021/acsmedchemlett.2c00427
- [21] Rutkowska E, Pajak K, Jozwiak K. Lipophilicity: methods of determination and its role in medicinal chemistry. Acta Pol Pharm 2013; 70(1):3–18.
- [22] Weni M, Safithri M, Seno DSH. Molecular docking of active compounds *Piper crocatum* on the alpha-glucosidase enzyme as anticiabetic. Indonesian J Pharm Sci Tech 2020; 7(2):64–72; https://doi.org/10.24198/ijpst.v7i2.21120
- [23] Lipinski CA, Lombardo F, Dominy BW, Feeney PJ. Experimental and computational approaches to estimate solubility and permeability in drug discovery and development settings. Adv Drug

- Delivery Rev 2001; 46(1-3):3-26; https://doi.org/10.1016/ S0169-409X(00)00129-0
- [24] Harper KC, Bess EN, Sigman MS. Multidimensional steric parameters in the analysis of asymmetric catalytic reactions. Nature Chem 2012; 4(1):366–74; https://doi.org/10.1038/nchem.1297
- [25] Sawale RT, Kalyankar TM, George R, Deosarkar SD. Molar refraction and polarizability of antiemetic drug 4-amino-5-chloro-N-(2-(diethylamino) ethyl)-2 methoxybenzamide hydrochloride monohydrate in (aqueous-sodium or lithium chloride) solutions at 30° C. J Appl Pharm Sci 2016; 6(3):120-4; https://doi.org/10.7324/JAPS.2016.60321
- [26] Geerts T, Heyden TV. In silico predictions of ADME-Tox properties: drug absorption. Comb Chem High Throughput Screen 2011; 14(5):339–61; https://doi.org/10.2174/138620711795508359
- [27] Kapetanovic IM. Computer-aided drug discovery and development (CADD): in silico-chemico-biological approach. Chem Biol Interact 2008; 171(2):165–76; https://doi.org/10.1016/j.cbi.2006.12.006
- [28] Ghannay S, Kadri A, Aouadi K. Synthesis, in vitro antimicrobial assessment, and computational investigation of pharmacokinetic and bioactivity properties of novel trifluoromethylated compounds using in silico ADME and toxicity prediction tools. Monatsh Chem 2020; 151:267–80; https://doi.org/10.1007/ s00706-020-02550-4
- [29] Srivalli KMR, Lakshmi PK. Overview of P-glycoprotein inhibitors: a rational outlook. Braz J Pharm Sci 2012; 48:353–67; https://doi.org/10.1590/S1984-82502012000300002
- [30] Morris-Schaffer K, McCoy MJ. A review of the LD50 and its current role in hazard communication. ACS Chem Health Saf 2021; 28(1):25–33; https://doi.org/10.1021/acs.chas.0c00096
- [31] Nurlela N, Awaluddin F, Batubara I, Wahyudi ST. Computational study of kaurene diterpenoids for antivirals against sarscov-2. J Appl Pharm Sci 2022; 12(8):112–29; https://dx.doi. org/10.7324/JAPS.2022.120812
- [32] Noga M, Michalska A, Jurowski K. Application of toxicology in silico methods for prediction of acute toxicity (LD50) for Novichoks. Arch Tox 2023; 97:1691–700; https://doi. org/10.1007/s00204-023-03507-2
- [33] Pires DEV, Blundell TL, Ascher DB. Pkcsm: predicting small-molecule pharmacokinetic and toxicity properties using graphbased signatures. J Med Chem 2015; 58:4066–72; https://doi. org/10.1021/acs.jmedchem.5b00104
- [34] Daina A, Michielin O, Zoete V. Swissadme: a free web tool to evaluate pharmacokinetics, drug-likeness and medicinal chemistry friendliness of small molecules. Sci Rep 2017; 7:42717; https://doi.org/10.1038/srep42717
- [35] Dey M, Li X, Kunz RÖ, Ragsdale SW. Detection of organometallic and radical intermediates in the catalytic mechanism of methyl-coenzyme M reductase using the natural substrate methyl-coenzyme M and a coenzyme B substrate analogue. Biochemistry 2010; 49:10902–11; https://doi.org/10.1021/bi101562m
- [36] Wongnate T, Ragsdale SW. The reaction mechanism of methyl-coenzyme m reductase. J Bio Chem 2015; 290(15):9322-34; https://doi.org/10.1074/jbc.M115.636761
- [37] Magalhães RP, Vieira T, Melo A, Sousa SF. In silico development of quorum sensing inhibitors. In: Simões M, Borges A, Simões LC (eds.). Recent trends in biofilm science and technology, Academic Press, London, UK, pp 329–57, 2020; https://doi.org/10.1016/ B978-0-12-819497-3.00015-5
- [38] Kusuma SMW, Utomo DH, Susanti R. Molecular mechanism of inhibition of cell proliferation: an in silico study of the active compounds in Curcuma longa as an anticancer. J Trop Biodivers Biotech 2022; 7(3):1–16; https://doi.org/10.22146/jtbb.74905
- [39] Amrulloh LSWF, Harmastuti N, Prasetiyo A, Herowati R. Analysis of molecular docking and dynamics simulation of mahogany (Swietenia macrophylla king) compounds against the plpro

- enzyme sars-cov-2. Pharm Pharm Sci J 2023; 10(3):347–59; https://doi.org/10.20473/jfiki.v10i32023.347-359
- [40] Nolling J, Elfner A, Palmer JR, Steigerwald VJ, Pihl TD, Lake JA, et al. Phylogeny of *Methanopyrus kandleri* based on methyl coenzyme M reductase operons. Int J Syst Bacteriol 1996; 46:1170–3; https://doi.org/10.1099/00207713-46-4-1170
- [41] Badraoui R, Saeed M, Bouali N, Hamadou WS, Elkahoui S, Alam MJ, et al. Expression profiling of selected immune genes and trabecular microarchitecture in breast cancer skeletal metastases model: effect of alpha-tocopherol acetate supplementation. Calcif Tissue Int 2022; 110:475–88; https://doi.org/10.1007/s00223-021-00931-3