

# Network Analysis, and Experimental Validation to Uncover the Mechanism of the Four Compounds in *Artemisia annua* (Qing Hao) Antimalarial

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## Research

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# Abstract

**Background:** Artemisinin is widely used to treat malaria, but the antimalarial mechanism and coordinative interactions governing the actions of artemisinin, scopoletin, arteannuin B and artemisinic acid have not been elucidated.

**Methods:** Based on the existence of antimalarial drugs, the antimalaria targets of artemisinin, scopoletin, arteannuin B and artemisinic acid were investigated by molecular docking using the similarity theory of chemical structure, and the antimalaria mechanism of scopoletin and its coordinative antimalaria interactions with the other three ingredients of the mixture were subsequently examined.

**Results:** Using the text information excavation method, the relevant proteins involved in the antimalarial effect of artemisinin were IL-6, ACHE, PC3, IPOB, CYC, TNF- $\alpha$ , UGT1A9, CASP3, XDH, IL-1 $\beta$ , VEGF, CAT, CREB, AMPK, UGT1A6, ADR, MAPK, COX2, LB24AB and CYP450. The relevant proteins involved in the antimalarial effect of scopoletin were TNF- $\alpha$ , PI3K, IL-8, IL-6, VEGF, IL-1 $\beta$ , MAPK, CD4, SP2, CTNNB, CASP3, PRO1400, IgE, IL-4, ICAM1, p38, STAT3, TLR4 and API4. However, arteannuin B and artemisinic acid had little relevance to the abovementioned proteins. The interaction property between TNF- $\alpha$  and *Artemisia annua* was that the effect of the mixture of artemisinin, scopoletin, arteannuin B and artemisinic acid was greater than that of artemisinin alone, and the synergistic effect of the four elements was considered beneficial to the progress of antimalarial treatment.

**Conclusion:** The antimalarial targets of *Artemisia annua* ingredients were examined using data mining methods, and the antimalarial effect of scopoletin may be related to TNF. The combined application of the four elements achieved the same antimalarial effect and reduced the clinical use of artemisinin and scopoletin.

## Background

Malaria is a major threat to human life. The World Health Organization (WHO) lists malaria, AIDS and cancer as the world's three major deadly diseases. Before the emergence and promotion of artemisinin, approximately 400 million people worldwide were infected with malaria, and at least 1 million people died from malaria annually. The morbidity and mortality of malaria are especially high in sub-Saharan Africa. According to WHO data for 2016, governments and international organizations spent 2.7 billion dollars on malaria control and elimination worldwide [1]. Antimalarial drugs primarily include quinolines, such as chloroquine, mefloquine, and quinine, and antifolates, such as pyrimethamine and sulfadoxine. The use of these drugs effectively controlled the global spread of malaria. However, *Plasmodium falciparum* developed resistance to almost all antimalarial drugs [2]. China was once most strongly affected by malaria and undertook large-scale efforts to eliminate the disease [3]. The unique chemical compositions of the traditional Chinese materia medica have significant biological activity in major diseases. Chinese medicines with clear active ingredients are rare, and *Artemisia annua* (Qing Hao) is a typical chemical composition obtained by the modern, scientifically verified traditional Chinese materia medica. Victory

over the disease was finally achieved in 2014, when the number of malaria patients was controlled to 56 individuals in China [4]. Artemisinin was widely used in Thailand and other countries in Southeast Asia in the 1990s. Artemisinin was widely adopted in Africa and the Americas [5]. However, neither artemisinin monotherapy nor artemisinin-based combination therapy was efficacious, and delayed parasite clearance often confused clinicians, similar to other antimalarial drugs [6-8]. *Plasmodium falciparum* developed resistance to artemisinin in the Greater Mekong region, including Cambodia, Laos, Myanmar, Thailand and Vietnam. The WHO 2011 global plan attempted to tackle artemisinin resistance caused by the artemisinin partner drug [9]. The primary reason for this resistance may be the artemisinin partner drug [10]. The combining of artemisinin with another drug with low drug resistance may not delay parasite clearance. It remains entirely possible to rely on artemisinin and its new partner drugs to end the prevalence of malaria [11]. Drugs with low drug resistance must be novel because existing antimalarial drugs developed strong resistance. Scopoletin, arteannuin B and artemisinic acid were considered because their chemical structures are more similar to artemisinin. Compounds that selectively act on two or more targets of interest would theoretically have more pharmacological actions than single-target agents [12]. 'Polypharmacology' is a new methodology in drug discovery [13]. The present study used the method of network pharmacology to perform text mining and target prediction of four components of artemisinin, artemisinin, artemisinic acid and sorghum lactone, which may have antimalarial effects in *Artemisia annua* L.

Molecular docking is an essential procedure to verify network pharmacology in structural molecular biology and computer-assisted drug design. Molecular docking may be used to perform virtual screening of chemical compounds, rank the results, and propose structural hypotheses on how the ligands inhibit the target, which is highly valuable in lead optimization [14].

## Materials And Methods

### Reagents and materials

1-Ethyl-3-(3-dimethylaminopropyl)-carbodiimide hydrochloride (EDC) was obtained from Shanghai Beinuo Biotechnology Co., Ltd. (Shanghai, China), and N-hydroxysuccinimide (NHS) was purchased from Shanghai Dibai Biotechnology Co., Ltd. Sodium acetate and dimethyl sulfoxide (DMSO) were purchased from Beijing Chemical Plant (Beijing, China). Ethanolamine hydrochloride was purchased from Shanghai Sanying Chemical Reagent Co., Ltd. (Shanghai, China). The protein TNF- $\alpha$  was provided by RD Biosciences (USA). Artemisinin, scopoletin, arteannuin B and artemisinic acid were provided by Chengdu Ruifensi Biological Technology Co., Ltd. (Chengdu, China), and Dulbecco's phosphate-buffered saline PBS buffer (pH 4.7) was freshly prepared.

### Plasma sample preparation

Approximately 40 mg of EDC and 10 mg of NHS were weighed, and 1 mL of solution was prepared with distilled water. This solution was injected within 5 min into two channels that were thoroughly rinsed with PBS buffer.

Fifty micrograms of TNF- $\alpha$  protein was dissolved in 100  $\mu$ L of PBS, and 10  $\mu$ L of this solution was taken in three portions and diluted with sodium acetate solutions with pH values of 5.5, 6.0, and 6.5, respectively, and the final concentration was 50  $\mu$ g/mL. The flow rate was reduced to 20  $\mu$ L/min, and the left channel was rinsed for 10 min to determine the optimal pH of sodium acetate.

After determination of the optimal pH value, 1 M ethanolamine hydrochloride was injected into the two channels for 10 min to complete the sample fixation.

The four chemical compounds, scopoletin (A), artemisinin (B), artemisinic acid (C) and arteannuin B (D), were divided into 12 groups according to the combination rule of A, B, C, D, AB, AC, AD, ABC, ABD, ACD, BCD, and ABCD, and each group had 6 concentration levels. The control group was PBS buffer at pH 7.4.

Next, 19.2 mg of *Scutellaria* lactone, 28.2 mg of artemisinin, 24.8 mg of artemisinin 2 and 23.4 mg of artemisinin were accurately weighed and dissolved in 1 mL DMSO (dimethyl sulfoxide). The DMSO solution in each group was gradient-diluted with PBS to final concentrations of 200  $\mu$ M, 66.7  $\mu$ M, 22.2  $\mu$ M, 7.41  $\mu$ M, and 2.47  $\mu$ M.

## Target fishing

Known therapeutic targets for the treatment of malaria were obtained from the DrugBank database (<http://www.drugbank.ca/>, version 4.3) [15]. The prediction of drug targets based on ligand structural features primarily includes chemical similarity searches and reverse pharmacophore searches. The theoretical basis of the chemical similarity search is that small molecular compounds with similar structural or physicochemical properties act on targets with the same or similar properties: “antimalaria” was selected as the key word, and the drug-target interactions of drugs approved by the USA Food and Drug Administration (FDA) for the treatment of menstrual disorders. All target gene/protein identifiers (IDs) were converted into the corresponding gene symbol/UniProtKB-Swiss-Prot IDs to facilitate further data analyses. After removing redundant entries, 25 target genes corresponding to 15 known antimalarial drugs were retrieved.

## Protein-protein interaction (PPI) data

PPI data were imported from the Human Annotated and Predicted Protein Interaction Database (HAPPI, <http://bio.informatics.iupui.edu/HAPPI/>, Version 31.2) [16]. Based on this PPI network database, an interaction network of *Artemisia annua* candidate target groups and known antimalarial drug target groups was constructed, and the distribution of target nodes in metabolic pathways and the corresponding diseases was determined. A direct interaction network of key nodes was established and divided into different modules according to the functions of the nodes. According to the malaria pathway (ko05144: Malaria) in the Kyoto Encyclopedia of Genes and Genomes (KEGG), molecules closely related to the malaria pathway were selected as candidates to be verified from the key nodes. The MCODE algorithm is used to intersect the PPI network for module analysis. The score was 2.2-4.7, while the node was 3-21, and the edge was 2-58.

3. The unit of docking score is kJ/mol. And the unit was mentioned in the manuscript.

### **Network construction and topological analysis**

Compound–target (C-T), target–pathway (T-P), and target–disease (T-D) networks of malaria were constructed using Cytoscape 3.2 software (<https://cytoscape.org/download.html>), which is a general bioinformatics software package for data integration and visualization of biological networks (Bindea *et al.*, 2009; Smoot *et al.*, 2011). An interaction network of *Artemisia annua* candidate target genes with known antimalarial drug target genes was established and consisted of 85 nodes and 298 pairs of interactions. The topological characteristic value of each node was calculated in the network, and the median of the topological characteristic value was used as the card value. A total of 32 key nodes were screened. A direct interaction network of key nodes was established and processed according to the node functions. The module was divided, the malaria pathway in KEGG (ko05144: Malaria) was compared, and molecules closely related to the malaria pathway were selected as candidates for verification from the key nodes.

### **Molecular docking**

The molecular structures of CDK4, NFKB1, PIK3CG, MAPK1, TNF and ITGB2 human protein targets were searched in the database UniProt (<http://www.uniprot.org/>). The structures of scopolamine and artemisinic acid were downloaded from the PubChem database (<https://pubchem.ncbi.nlm.nih.gov>). Chemical compositions and protein structures were dehydrated and hydrotreated, respectively. Molecular docking and figures were generated using Discovery Studio Visualizer 2.5 software.

### **Probe Kd determination**

Approximately 40 mg of EDC and 10 mg of NHS were weighed, and 1 mL of solution was prepared with distilled water. This solution was injected within 5 min into two channels that were thoroughly rinsed with PBS buffer. Next, 19.2 mg of scopolamine, 28.2 mg of artemisinin, 24.8 mg of artemisinin, and 23.4 mg of artemisinic acid were precisely weighed in 1 mL DMSO and mixed well. TNF- $\alpha$  protein was immobilized on grafted sensor chips. The compound monomers and combinations were divided into 12 groups (Table 1). Each group of samples was injected from low to high concentrations, and a control group (PBS) was set at each concentration. Regression analysis was performed when concentration curves were separated and approaching equilibrium. The dissociation constant (Kd) and its maximum value (Bmax) were calculated by fitting the titration curve to the single-site saturation binding equation  $[Y = B_{max} * X / (K_d + X)]$  using GraphPad Prism software (GraphPad Software Incorporated, La Jolla, CA, USA).

Table 1 Compound monomers and combinations

Group	Compounds
A	Scopoletin (C <sub>10</sub> H <sub>8</sub> O <sub>4</sub> )
B	Artemisinin (C <sub>15</sub> H <sub>22</sub> O <sub>5</sub> )
C	Artemisinic acid (C <sub>15</sub> H <sub>22</sub> O <sub>2</sub> )
D	Arteannuin B (C <sub>15</sub> H <sub>20</sub> O <sub>3</sub> )
AB	Scopoletin; Artemisinin
AC	Scopoletin; Artemisinic acid
AD	Scopoletin; Arteannuin B
ABC	Scopoletin; Artemisinin; Artemisinic acid
ABD	Scopoletin; Artemisinin; Arteannuin B
ACD	Scopoletin; Artemisinic acid; Arteannuin B
BCD	Artemisinin; Artemisinic acid; Arteannuin B
ABCD	Scopoletin; Artemisinin; Artemisinic acid; Arteannuin B

## Results

### Compound–target network construction

There were 162 diseases related to the four chemical components observed in *Artemisia annua*. Artificial noise reduction identified the top 5 diseases, ranked by frequency of occurrence, as malaria (50), cerebral malaria (7), falciparum malaria (4), visceral leishmaniasis (2), and systemic lupus erythematosus (1) (Figure 1).

Fifteen known antimalarial drugs were retrieved from the DrugBank database (Table 2). There were 93 targets with structures similar to scopolide and 15 known antimalarial drugs (similar score greater than 0.7). Artemisinic acid had 32 targets with structures similar to 15 known antimalarial drugs (similar score greater than 0.7).

### Clustering analysis

The key node interaction network was divided into three functional modules. The first functional module primarily involved immune-related pathways, the second module primarily involved multiple infectious diseases, and the third module was related to drug metabolism and tumour pathways. The KEGG malaria pathway (ko05144: Malaria) comparison is shown in Figure 2. The four key nodes were involved in four

pathways: the T cell receptor signalling pathway, the Toll-like receptor signalling pathway, the TNF signalling pathway, and natural killer cell-mediated cytotoxicity, which all have important links to the malaria pathway. Therefore, the joint nodes involved in these four pathways were regarded as candidates for verification.

Table 2 Known antimalarial drugs

Drug ID	Name
DB00254	Doxycycline
DB00908	Quinidine
DB00358	Mefloquine
DB01611	Hydroxychloroquine
DB00608	Chloroquine
DB00613	Amodiaquine
DB01103	Quinacrine
DB01087	Primaquine
DB00468	Quinine
DB04877	Voacamine
DB00440	Trimethoprim
DB01218	Halofantrine
DB00205	Pyrimethamine
DB01131	Proguanil
DB01117	Atovaquone

## Molecular docking

Scopolide had suitable docking sites for TNF, NFKB1, and PIK3CG, and the docking scores were 77.9576, 57.8491, and 50.2248, respectively, with no suitable docking site being identified for CDK4 or MAPK1. There was no suitable docking site for artemisinic acid with ITGB2 (Table 3, Figure 3).

Table 3 Docking score of scopolide and artemisinic acid with the targets

Compounds	Targets	Docking Score (kJ/mol)
Scopolide	CDK4	-
Scopolide	NFKB1	57.8491
Scopolide	PIK3CG	50.2248
Scopolide	MAPK1	-
Scopolide	TNF	77.9576
Artemisinic acid	ITGB2	-

### Experimental validation of key targets

According to the binding curve of the immobilized protein, the highest binding efficiency of sodium acetate solution at pH 5.5 was 110  $\mu$ RIU/s, but the binding was unstable. The lowest binding efficiency at pH 6.5 was 68  $\mu$ RIU/s. The binding efficiency was better at pH 6.0, and the comparison of binding was stable at 108  $\mu$ RIU/s. Therefore, the optimal approach for protein fixation was sodium acetate solution at pH 6.0.

TNF had good binding with scopolide, which suggests that the antimalarial effect of scopolide may be related to TNF. The binding of TNF to artemisinin, artemisinin B, and artemisinic acid was weak, but the combination of the 4 components of artemisinin, artemisinin B, artemisinic acid, and stigmalactone had good binding to TNF, which suggests that the combined application of 4 ingredients may achieve antimalarial effects by acting on TNF (Table 4, Figure 4).

Table 4 Equilibrium dissociation constant (Kd) for each group

Group	Kd (M)	Est. error	Bmax Signal (mRIU)
A	$4.05 \times 10^{-6}$	$2.00 \times 10^{-6}$	42.0
B	$2.01 \times 10^{-4}$	$1.45 \times 10^{-5}$	38.7
C	$8.91 \times 10^{-5}$	$4.07 \times 10^{-7}$	72.3
D	$2.93 \times 10^{18}$	$1.39 \times 10^{-7}$	$2.78 \times 10^{17}$
AB	$1.35 \times 10^{-3}$	$8.08 \times 10^{-6}$	46.7
AC	$1.25 \times 10^{-4}$	$3.97 \times 10^{-6}$	33.7
AD	$3.29 \times 10^{-4}$	$1.72 \times 10^{-6}$	15.4
ABD	$1.20 \times 10^{-4}$	$1.61 \times 10^{-6}$	12.7
ABC	$2.69 \times 10^{-4}$	$8.38 \times 10^{-7}$	25.8
ACD	$1.59 \times 10^{20}$	$4.91 \times 10^{-7}$	$8.36 \times 10^{18}$
BCD	$2.80 \times 10^{-5}$	$1.78 \times 10^{-5}$	10.1
ABCD	$2.17 \times 10^{-5}$	$5.04 \times 10^{-5}$	12.5

## Discussion

The sesquiterpene compounds represented by artemisinin are the most heavily researched compounds in *Artemisia annua*. Nearly 61 sesquiterpenoids, primarily artemisinin compounds, were identified from *Artemisia annua*, including artemisinic acid, artemisinol, artemisinin ether and artemisinin. Artemisinin is a sesquiterpene lactone that contains an endoperoxide bridge structure, and it is the main component of several antimalarial treatments. Research data showed that artemisinin and artemisinin may be converted to artemisinin in the original plants [17, 18]. The biosynthetic routes of artemisinin may be summarized as 9 total synthetic routes and 5 semi-synthetic routes. The synthesis method of artemisinin is a long process with high cost and low total output that produces a maximum yield of 10%. Artemisinic acid is one of the main components of sesquiterpenes in *Artemisia annua* plants and is an important precursor of artemisinin synthesis. Tu Yu showed that young plants of *Artemisia annua* contained a large amount of artemisinic acid but exhibited a shortage of artemisinin. It is speculated that sesquiterpenoids, such as artemisinin, are converted from artemisinic acid. Levesque F and other researchers used synthetic biology to successfully produce artemisinin using genetically engineered yeast [19]. Scopolide has strong water solubility and stability to artemisinin and has pharmacological activity that reflects the efficacy of traditional artemisinin. Previous studies showed that scopolide had certain antimalarial

effects and certain synergistic effects with artemisinin. With the rapid development of chemical genomes and pharmacological technologies, a large number of potential targets and massive biological activity data emerged. However, with the accumulation of complex data, simple analysis methods no longer satisfy the analytical needs of high-throughput and large-scale data [20]. The rapid development of chemical informatics recently met the requirements of big data processing and information extraction tasks that are urgently needed in chemical genomics. Chemical informatics primarily studies how to properly select compounds from diverse compound libraries, how to describe drug molecular characteristics, how to measure the differences between different molecules, how to identify drug-like molecules, molecular structure and biological performance relationships, and how to develop corresponding computer software and hardware. This methodology includes the research tasks and content of chemometrics and computational chemistry [21]. One important application of the chemoinformatics method in the post-genomic era is predicting the potential targets of small-molecule compounds based on existing biological and chemical information and explaining their mechanism of action to accelerate the development of drugs. The prediction of drug targets is considerably important to the evaluation of early drug molecules and the new use of old drugs. However, due to the limitations of throughput, accuracy and cost, it is difficult to widely apply experimental methods. As a quick and low-cost method, the development of computer-aided target prediction algorithms is receiving increasing attention. According to different research strategies, the prediction of drug targets based on chemoinformatics may be divided into three categories: predictions based on ligand characteristics; predictions based on protein structural characteristics; and predictions based on data mining methods [22, 23]. According to the prediction of targets, the potential targets of scopolide are CDK4, NFKB1, PIK3CG, MAPK1, and TNF, and the potential target of artemisinic acid is ITGB2. Cyclin-dependent kinase (CDK) is a serine/threonine (Thr) kinase that is an important signal transduction molecule in cells. CDK-cyclin is formed by the cyclin complex and is involved in cell growth, proliferation, dormancy, or apoptosis. During the cell cycle, cyclins are periodically and continuously expressed and degraded and bound to CDKs, which are transiently activated by cyclins. Cyclins catalyse the phosphorylation of different substrates via CDK activity to promote and transform different phases of the cell cycle. The CDK family includes CDK1-13, and cyclins are divided into cyclins A-L. Different CDKs are connected to different cyclins. CDK4/6-specific activation is closely related to the proliferation of some tumours. Rb is present in approximately 80% of human tumours, and abnormalities in the cyclin D–CDK4/6–INK4–Rb pathway are common [24]. It is characterized by (1) p16<sup>INK4a</sup> gene deletion, point mutation, or DNA methylation, which leads to the inactivation of p16<sup>INK4a</sup>, and (2) CDK4 gene amplification or point-mutated T cells induce other cells to activate or interfere with lysis. CD3, CD4, and CD8 cells are involved in the T cell transcription of activation signals. Toll-like receptors (TLRs) play an important role in the identification of invading pathogenic microorganisms in early congenital immunity. These evolutionarily preserved receptors are homologous to the Drosophila Toll protein family in structure, and they recognize highly conserved structural motif (motif)-pathogen-associated molecular patterns expressed only on pathogenic microorganism molecular patterns (PAMPs). PAMPs stimulate TLRs to initiate a signalling cascade that includes several proteins that lead to the activation of the transcription factor NF- $\kappa$ B, which induces the secretion of pro-inflammatory cytokines and effector cytokines that are directly involved in

the adaptive immune response. Integrin  $\beta 2$  (CD18) is an important member of the integrin family of adhesion molecules. Integrin  $\beta 2$  binds to different integrin subunits to form the leukocyte adhesion receptor group. Integrin  $\beta 2$  is primarily expressed in white blood cells, and its ligands are TCAM, iC3b, and fibrinogen. The cytoplasmic region of integrin  $\beta 2$  is linked to a variety of cytoskeletal proteins and is involved in signal transduction. The genetic defects of integrin  $\beta 2$  lead to leukocyte adhesion deficiency syndrome. Integrin  $\beta 2$  primarily exists in the natural killer cell-mediated cytotoxicity pathway. The present study used the biological macromolecule interaction instrument to verify the binding of the target protein TNF with scopolide, which had the highest score in the docking experiment. The binding effect was good and confirmed that scopolide acted on TNF and participated in its corresponding functions. The results confirmed that scorolactone had an antimalarial effect. However, the binding rates of artemisinin, artemisinin B, artemisinic acid and TNF were very low, which may be observed because these three compounds did not act on TNF. This finding also showed that the mechanism of the antimalarial effect of artemisinin may be different. The combination of the four ingredients with TNF was very high, which indicates that the four ingredients have a synergistic effect. The combined use of the four ingredients may reduce the amount of artemisinin and scopolamine used to achieve the same antimalarial effect. The characteristics of multicomponent, multitarget, and synergistic effects are common in traditional Chinese medicine preparations. Many experiments indicated that TNF had a certain killing effect on Plasmodium. TNF must be assisted by a certain factor (or factors) in the body to exert its ability to damage Plasmodium, which means TNF is not a terminal effector that directly kills Plasmodium. The immune-protective mechanism of TNF may have the following effects. (1) Enhancing the phagocytosis function of phagocytic cells: one study treated neutrophils with different doses of TNF for 30 min followed by incubation with *Plasmodium falciparum* and found that the phagocytosis of each stage of Plasmodium was strengthened, and the extent of the increase positively correlated with the dose of TNF within a certain range. (2) Acting via reactive oxygen mediator: when TNF and macrophages were co-incubated for 30 min, the release of reactive oxygen species (ROS) from macrophages was detected, and ROS kill Plasmodium.

## Conclusion

Based on the existing literature, data mining methods were used to identify the targets of the antimalarial active ingredients artemisinin and scopolamine in *Artemisia annua*, including 20 proteins related to antimalarial actions of artemisinin (IL-6, ACHE, PC3, IPOB, CYC, TNF- $\alpha$ , UGT1A9, CASP3, XDH, IL-1 $\beta$ , VEGF, CAT, CREB, AMPK, UGT1A6, ADR, MAPK, COX2, LB24AB, and CYP450) and 19 proteins related to the antimalarial properties of scopolamine (TNF- $\alpha$ , PI3K, IL-8, IL-6, VEGF, IL-1 $\beta$ , MAPK, CD4, SP2, CTNNB, CASP3, PRO1400, IgE, IL-4, ICAM1, P38, STAT3, TLR4, API4). No protein targets of artemisinic acid or arteannuin B were identified.

The combination of the four components artemisinin, arteannuin B, artemisinic acid, and stigmalactone exhibited good binding to TNF.

# Abbreviations

NHS: N-hydroxysuccinimide; DMSO: dimethyl sulfoxide; PPI: protein-protein interaction; KEGG: Kyoto Encyclopedia of Genes and Genomes; Thr: threonine; CDK: cyclin-dependent kinase; PAMP: pathogenic microorganism molecular pattern.

# Declarations

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## Authors' contributions

Z-yP, W-L, and H-IQ integrated the pharmacokinetics study. Z-xB and X-hY performed the network analysis and experimental validation to research the antimalaria mechanisms of artemisinin, scopoletin, arteannuin B and artemisinic acid. Z-yP and H-IQ wrote the manuscript.

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## Availability of data and materials

All relevant data are included in this report.

## Ethics approval and consent to participate

There are no ethics statement in this paper.

## Consent

Not applicable.

## Competing interests

The authors declare that they have no competing interests.

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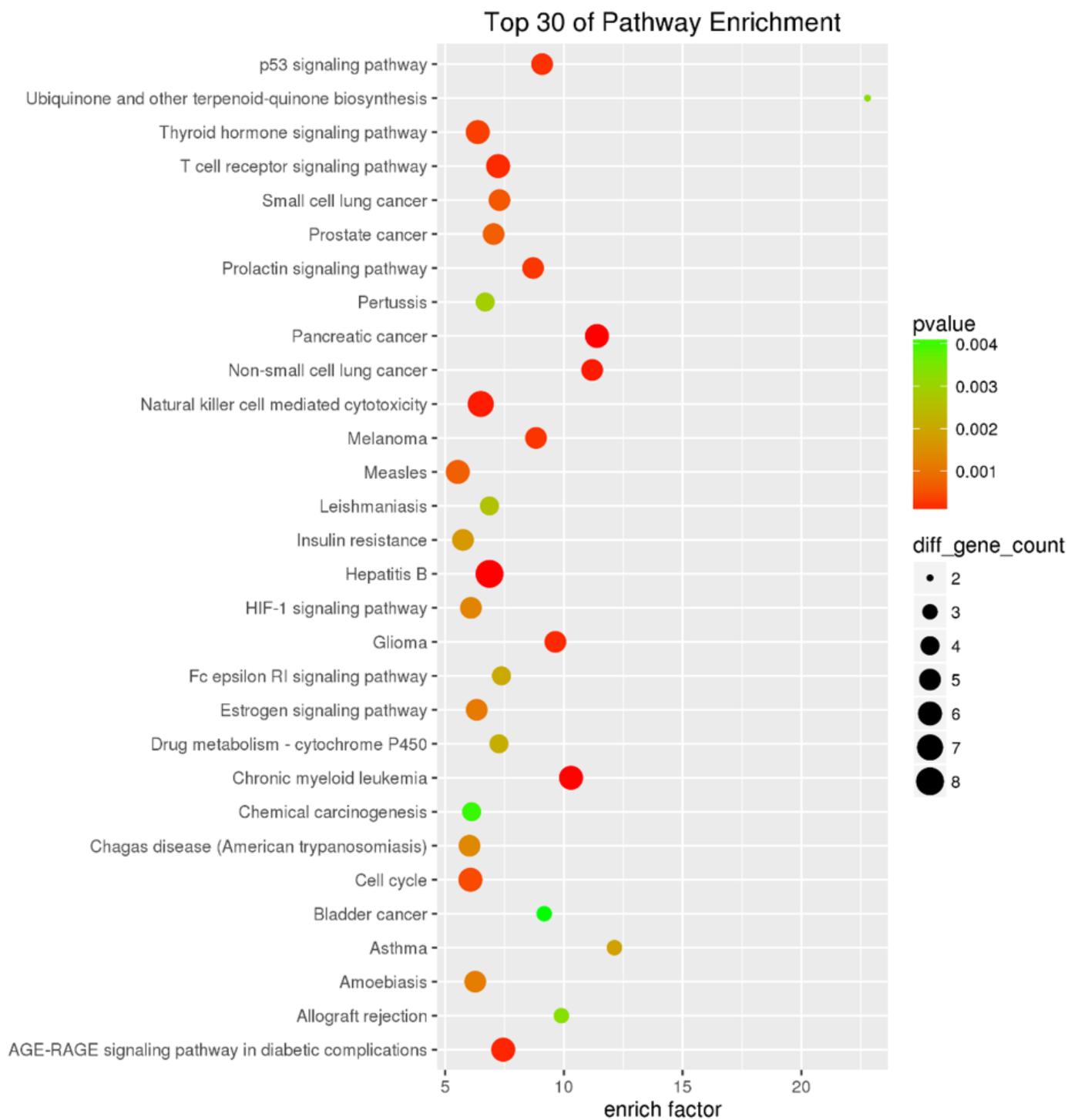
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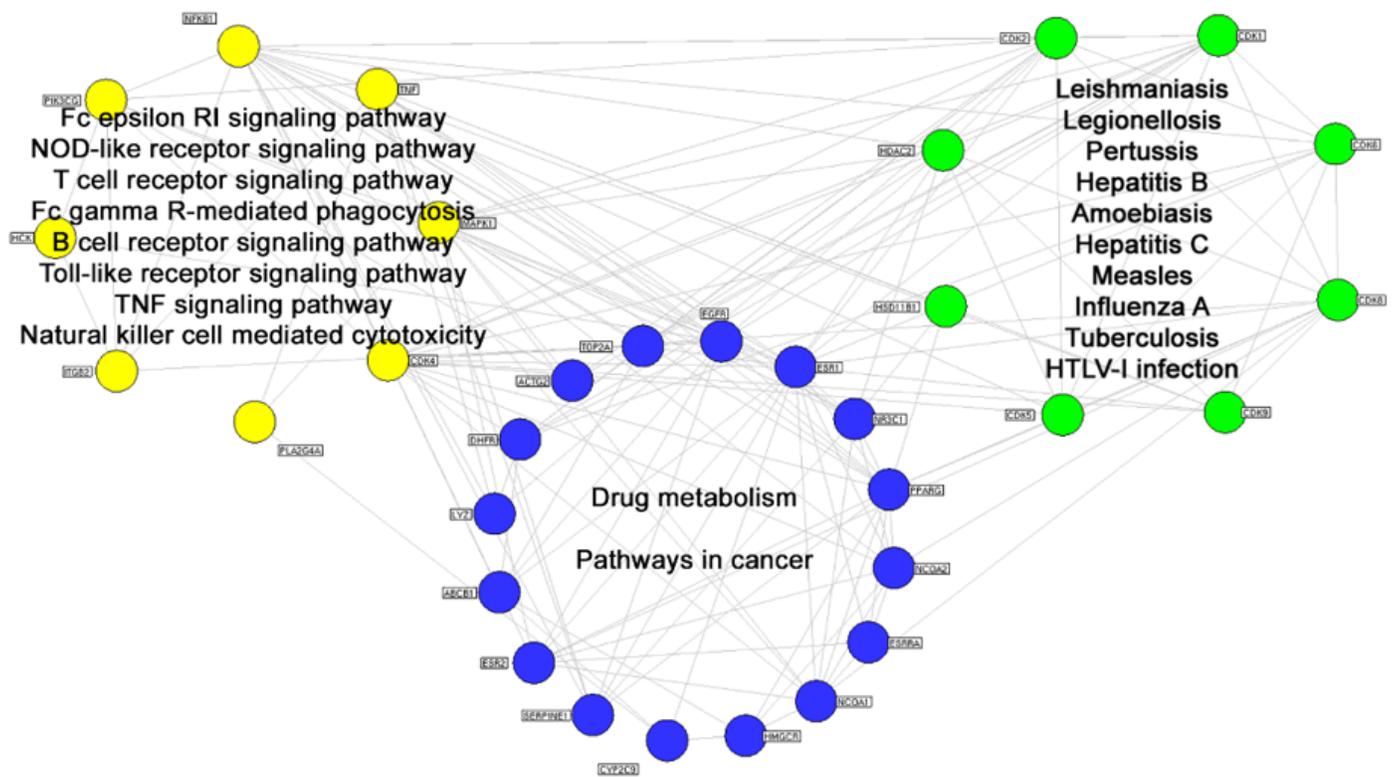
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## Figures



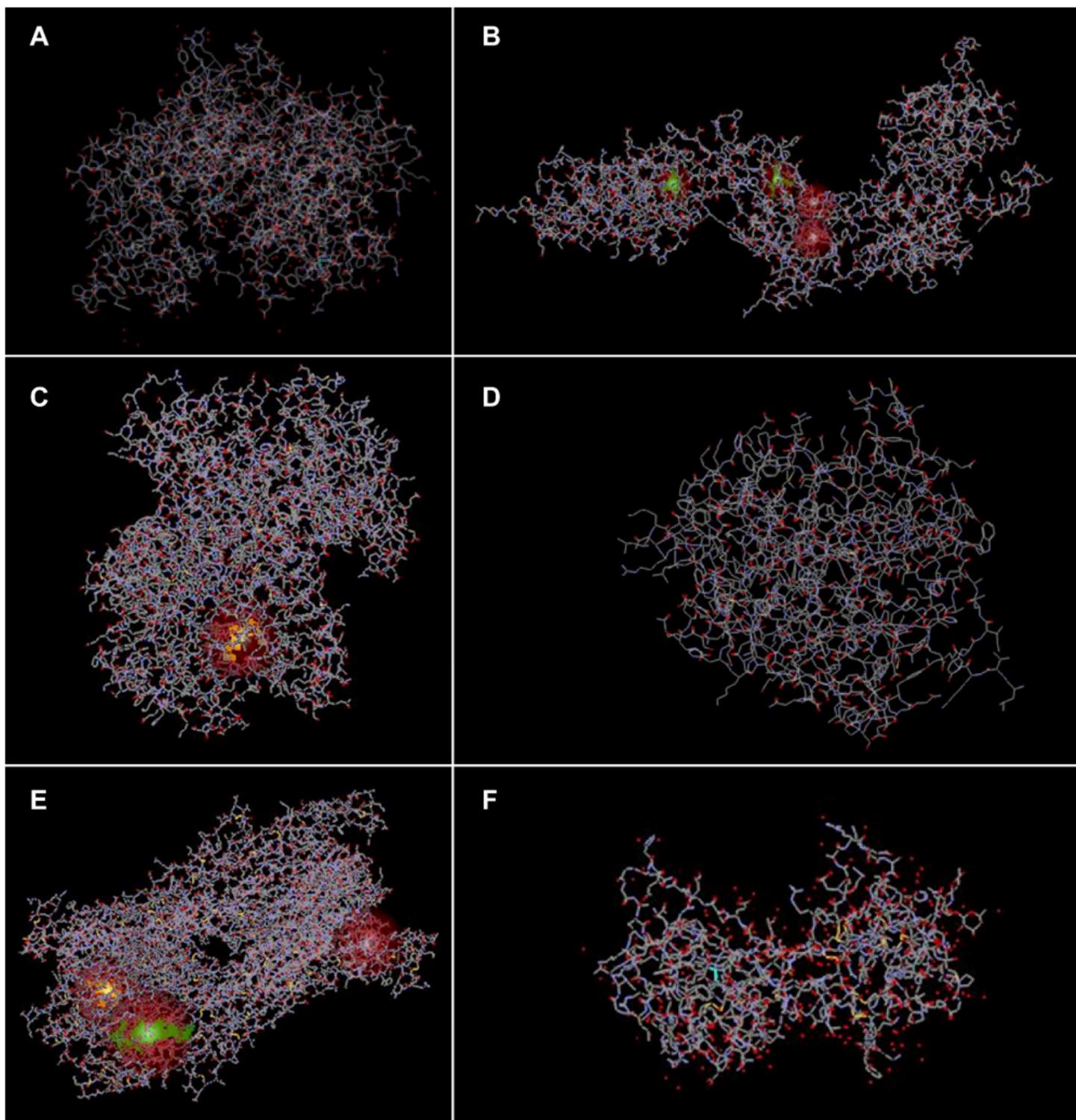
**Figure 1**

The above candidate targets significantly participate in multiple immune-related pathways, such as Fc epsilon RI signaling pathway, T cell receptor signaling pathway, and Natural killer cell mediated cytotoxicity; and multiple infectious disease-related pathways, such as Hepatitis B (hepatitis B), Leishmaniasis (leish (Mann disease), Pertussis, Amoebiasis, Measles, Toxoplasmosis and Hepatitis C.



**Figure 2**

Establishment of an interaction network between *Artemisia annua* candidate target genes and known antimalarial drug target genes, consisting of 85 nodes and 298 pairs of interactions.



**Figure 3**

Molecular Docking by Discovery Studio. A Scopolide with CDK4; B Scopolide with NFKB1; C Scopolide with PIK3CG; D Scopolide with MAPK1; E Scopolide with TNF; F Artemisinic acid with ITGB2.

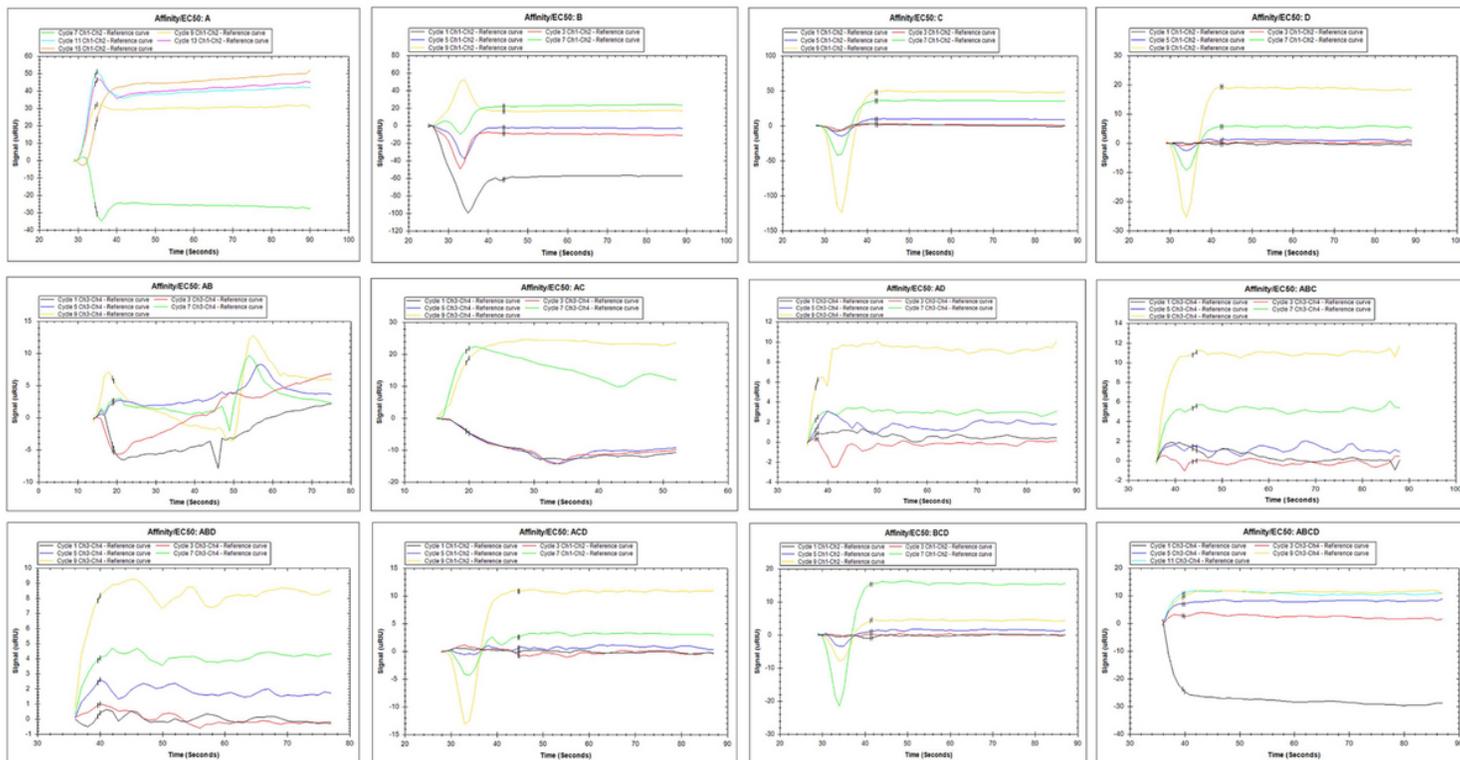


Figure 4

The dissociation constant (Kd) of each group. Grouping is the same as in Table 1.