

Identification of the active compounds and significant pathways of yinchenhao decoction based on network pharmacology

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Abstract. Yinchenhao decoction (YCHD) is a traditional Chinese medicine formulation, which has been widely used for the treatment of jaundice for 2,000 years. Currently, YCHD is used to treat various liver disorders and metabolic diseases, however its chemical/pharmacologic profiles remain to be elucidated. The present study identified the active compounds and significant pathways of YCHD based on network pharmacology. All of the chemical ingredients of YCHD were retrieved from the Traditional Chinese Medicine Systems Pharmacology database. Absorption, distribution, metabolism and excretion screening with oral bioavailability (OB) screening, drug-likeness (DL) and intestinal epithelial permeability (Caco-2) evaluation were applied to discover the bioactive compounds in YCHD. Following this, target prediction, pathway identification and network construction were employed to clarify the mechanism of action of YCHD. Following OB screening, and evaluation of DL and Caco-2, 34 compounds in YCHD were identified as potential active ingredients, of which 30 compounds were associated with 217 protein targets. A total of 31 significant pathways were obtained by performing enrichment analyses of 217 proteins using the JEPETTO 3.x plugin, and 16 classes of gene-associated diseases were revealed by performing enrichment analyses using Database for Annotation, Visualization and Integrated Discovery v6.7. The present study identified potential active compounds and significant pathways in YCHD. In addition, the mechanism of action of YCHD in the treatment of various diseases through multiple pathways was clarified.

Introduction

Yinchenhao decoction (YCHD) is a classical traditional Chinese medicine (TCM) formulation. YCHD has been used widely for the treatment of Yang jaundice and liver disorders. YCHD is composed of three Chinese medicinal herbs: *Artemisiae scopariae* herba (ASH, Yinchen), *Radix et Rhizoma Rhei* (RERR, Dahuang) and *Gardeniae Fructus* (GF, Zhizi). Pharmacologic studies have shown that this formulation can also be used to treat pancreatic carcinoma (1), liver injury (2,3), liver fibrosis (4), liver cirrhosis (5,6), nonalcoholic steatohepatitis (7), cholestasis (8) and diabetes mellitus (DM) (9).

In recent years, TCM monomers and TCM compounds have been studied extensively worldwide. Liu *et al* (10) predicted the molecular targets of YCHD based on systems-biology methods using the TCMGeneDIT database. However, only 17 main compounds were analyzed, and no active component in GF was identified (11). Other studies have focused only on the molecular mechanism of a certain aspect of YCHD, for example, immunity and metabolism, transport, signal transduction, and cell growth/proliferation (12). Previously, we found that genipin, a single component of GF, had inhibitory effects on human hepatocellular carcinoma cells (13). However, the active substances of YCHD and its specific molecular mechanism of action in the diseases mentioned above are not clear.

Any TCM formulation is a complex system with multiple components, multiple targets, and synergistic interactions among its components (14). Because of its complex chemical composition, it is extremely difficult to study its role in the body as a mixture. The complexity of TCM formulations makes their in-depth study difficult, whereas systems pharmacology provides new ideas and perspectives for the study of Chinese herbal compounds. Studies on the active substances of TCM formulations, identification of the targets of active components, and determination of the relationship between efficacious substances and diseases using systems pharmacology (15) and network pharmacology (16) can help elucidate the molecular mechanism of action of TCM formulations.

For Chinese herbal compounds administered via the oral route, the ingredients in a TCM formulation must first overcome the barriers posed by ADME (absorption, distribution, metabolism and excretion) processes, and only the molecules

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that pass through the barriers may be classed as ‘active molecules’ (17). These molecules bind to the targets in the body, thereby eliciting their actions. Then, drugs interact with the human body at the network level, as well as the overall level of the organ.

Therefore, based on analyses of ADME-related properties, identification of the active molecules in TCM formulations that pass across the body barrier and prediction of the network targets of active substances was undertaken. Thereafter, studies on the overall effect on the body, as well as the mechanism of action, was carried out. This strategy could provide a basis for in-depth understanding of the mechanism of action of TCM formulations. The workflow of the network-pharmacology approach in the present study is illustrated in Fig. 1.

Materials and methods

Identification of candidate compounds. All compounds of the three Chinese medicinal herbs in YCHD were collected from the Traditional Chinese Medicine Systems Pharmacology Database and Analysis Platform (TCMSP) database. The TCMSP database consists of 500 Chinese herbal medicines registered in the *Chinese Pharmacopoeia* (2010 edition) with 30,069 ingredients through literature mining and database integration. Data relevant to the pharmacokinetic properties of each chemical compound, which contained the prediction of oral bioavailability (OB), intestinal epithelial permeability (Caco-2 cells), drug-likeness (DL), blood-brain barrier (BBB), drug half-life (HL) and Lipinski's rule (LR) of five, were provided for the screening and evaluation of compounds (17).

Screening of active compounds. In ADME processes, OB is one of the most important pharmacokinetic parameters (18). High OB is often a key indicator to determine the DL of bioactive molecules. For TCM formulations, the failure of most of the ingredients to reach the protein target sites of particular cells is due to a lack of appropriate pharmacologic properties, especially OB. Molecules with OB $\geq 30\%$ were considered to have good OB in the present study.

In the early stages of drug development, DL evaluation helps to screen out excellent compounds (19) and increases the ‘hit rate’ of drug candidates. Therefore, the DL of molecules in YCHD was assessed using the Tanimoto coefficient in the present study (20) using the following formula:

$$T(X,Y) = \frac{x * y}{x^2 + y^2 - x * y}$$

Where x is the molecular descriptor of YCHD based on Dragon software (http://www.talet.mi.it/products/dragon_description.htm) and y is the average descriptor of all drugs in the Drugbank database. The average DL Index of all drugs in the Drugbank database is 0.18, which indicates a high DL. Thus in our study, active molecules were defined as those with a DL Index ≥ 0.18 .

The intestinal epithelial permeability can be investigated using Caco-2 cells (21). Orally administered drugs are absorbed mainly through intestinal epithelial cells. Therefore, simulation of drug transport across the monolayers of small-intestinal epithelial cells is crucial for the prediction of drug absorption. The permeability of epithelial cells of ingredients in Chinese

herbal medicines was predicted using the TCMSP database. It was considered that molecules with Caco-2 > -0.40 had good permeability in the small-intestinal epithelium.

Hence, the selected candidate molecules had to meet the requirements of OB $\geq 30\%$, DL ≥ 0.18 and Caco-2 > -0.40 for further analyses.

Identification of associated proteins and gene names. Protein targets were retrieved from the TCMSP database (<http://lsp.nwsuaf.edu.cn/tcmsp.php>). The dataset used in model-building comprised 6511 drug molecules and 3987 targets for which the compound-protein interactions are known in the Drugbank database (17). UniProt Knowledgebase (UniProtKB) is a protein database containing 54,247,468 sequence entries. The gene names were extracted further from the UniProtKB (<http://www.uniprot.org>).

Identification of significant pathways and gene-associated diseases. Java Enrichment of Pathways Extended to Topology (JEPETTO) is a Cytoscape 3.x plugin that performs integrative analyses of human gene sets. It can also identify functional associations between genes and known cellular pathways and processes using protein-interaction networks and topologic analyses (22). Significant pathways can be identified by enrichment analyses of proteins using JEPETTO. Analyses of gene-associated diseases were performed with acquired genetic information by the Database for Annotation, Visualization and Integrated Discovery (DAVID) v6.7 (23).

Construction of a network and analyses. The Compound-Target Network was built by connecting the candidate compounds and corresponding targets. The Compound-Pathway Network was generated by linkage of the candidate compounds and the signaling pathways involved. In the Gene-Disease Network, diseases were connected with the associated candidate targets. The corresponding diseases of potential genes were collected by DAVID enrichment analyses, and the obtained interactions between diseases and genes were applied further for building the Gene-Disease Network.

In this bilateral network, the ‘nodes’ represented the compounds, protein targets, signal pathways or diseases, and ‘edges’ represented the interactions of Compound-Target, Compound-Pathway or Gene-Disease. The networks were constructed using Cytoscape v3.3.0 (24).

Results

Identification of the active compounds in YCHD. Using the TCMSP database, 236 compounds were retrieved: 53 in ASH, 92 in RERR, and 98 in GF (3 herbs shared 7 compounds). The network flowchart of the compounds in YCHD is shown in Fig. 1. Of the 53 compounds in ASH, 34 satisfied the criterion of OB $\geq 30\%$, and 13 satisfied the criteria of OB $\geq 30\%$, DL ≥ 0.18 and Caco-2 ≥ -0.4 . Of the 92 compounds in RERR, 26 satisfied the criterion of OB $\geq 30\%$, 16 satisfied the criteria of OB $\geq 30\%$ and DL ≥ 0.18 , and 9 satisfied the criteria of OB $\geq 30\%$, DL ≥ 0.18 and Caco-2 ≥ -0.4 . Of the 98 compounds in GF, 43 satisfied the criterion of OB $\geq 30\%$, and 15 satisfied the criteria of OB $\geq 30\%$ and DL ≥ 0.18 , and 14 satisfied the criteria of OB

Table I. The number of compounds in YCHD satisfy $OB \geq 30\%$, $DL \geq 0.18$ and $Caco-2 \geq 0.4$

Herbs	Total	$OB \geq 30\%$	$DL \geq 0.18$	$Caco-2 \geq 0.4$
ASH	53	34 (64.2)	13 (24.5)	13 (24.5)
RRER	92	26 (28.3)	16 (17.4)	9 (9.8)
GF	98	43 (48.9)	15 (15.3)	14 (14.3)

YCHD, yinchenhao decoction; OB, oral bioavailability; DL, drug-likeness.

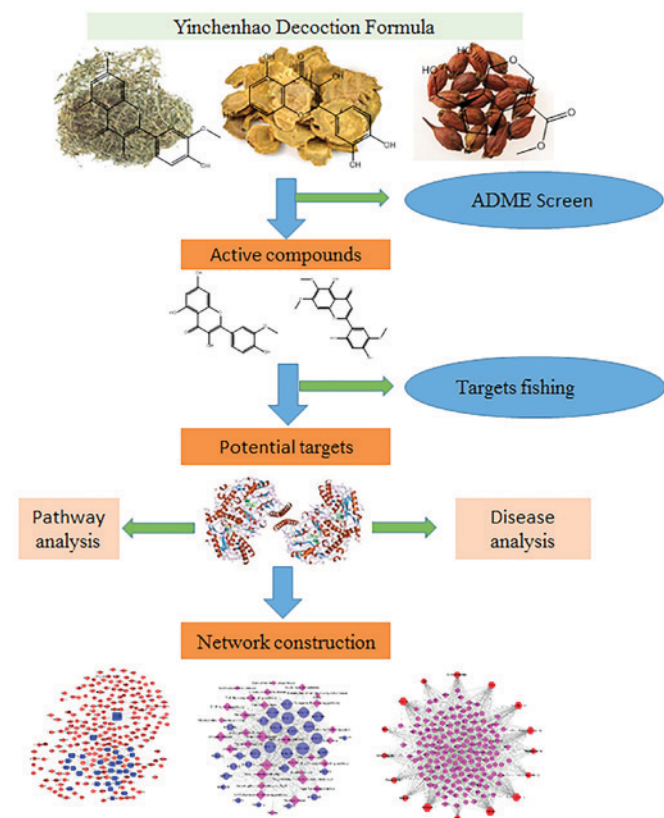


Figure 1. The workflow for the network-pharmacology approach used in our study.

$\geq 30\%$, $DL \geq 0.18$ and $Caco-2 \geq 0.4$. Among 243 compounds, 36 compounds satisfied all of the pre-defined requirements (Table I) and, finally, 33 compounds were analyzed after removing duplicates. The OB of genipin was $<30\%$, but it was a common compound in GF and was shown to have inhibitory effects on human hepatocellular carcinoma cells in our previous study (13). Hence, genipin was also regarded to be a candidate compound. The details of 34 compounds are shown in Table II. Interestingly, all three Chinese medicinal herbs in YCHD (i.e., ASH, RERR and GF) contained beta-sitosterol, whereas RERR and GF contained quercetin. Beta-sitosterol and quercetin are the common chemicals found in 188 TCM formulations according to the TCMSP database.

Identification of targets in YCHD. Among the 34 compounds obtained, 618 proteins and genes were obtained for

30 compounds, and 217 proteins and genes were included after removing duplicates. The 30 candidate compounds and all of the potential targets were applied to produce a plot of Compound-Target interactions, including 247 nodes (30 compounds and 217 targets) and 618 edges (Fig. 2). In Fig. 2, the red nodes are drug targets and the blue nodes are compounds, and the edges represent the interactions between them. The centralization and heterogeneity of the network was 0.603 and 2.355, respectively. This finding indicated that some nodes were more concentrated in the network than others. That is, the Compound-Target space was biased towards certain compounds and targets. As depicted in Fig. 2, MOL053 (quercetin) displayed the most target interactions (degree=154), followed by MOL173 (kaempferol, degree=63), MOL015 (beta-sitosterol, degree=38), MOL014 (isorhamnetin, degree=37) and MOL174 (stigmaterol, degree=31).

A TCM formulation is a complex system with various components; one component may act on multiple targets and display synergistic effects to treat diseases. These compounds with high degree nodes may perform important roles in the pharmacologic effect of YCHD. Protein targets acting as 'hubs' in the network were prostaglandin G/H synthase 2 (PTGS2; 28 interactions), heat-shock protein HSP 90 (20 interactions), prostaglandin G/H synthase 1 (19 interactions), nuclear receptor coactivator 2 (18 interactions), dipeptidyl peptidase IV (18 interactions) and mRNA of PKA catalytic subunit C-alpha (17 interactions).

Revealing the significant pathways. Enrichment analyses of 217 proteins were done using JEPETTO and 31 significant pathways were obtained (Table III). This XD-score is relative to the average distance to all pathways and represents a deviation (positive or negative) from the average distance. The q-value determines the significance of the overlap (Fisher's exact test) between the input information and the pathways. The Overlap/Size shows the number of overlapping proteins compared with the size of the pathway.

Enrichment algorithm analyses of the XD-score and q-value revealed the highest XD-score to be 1.47032, and the threshold value of XD-score in our study was 0.35. Eighteen disease pathways included 11 cancer pathways (non-small-cell lung cancer, small-cell lung cancer, bladder cancer, prostate cancer, endometrial cancer, colorectal cancer, glioma, pancreatic cancer, chronic myeloid leukemia, acute myeloid leukemia, melanoma) and one immune system-disease pathway (graft-vs.-host disease), three infectious disease-related pathways (leishmaniasis, malaria, Chagas disease), two neurodegenerative-disease pathways (prion diseases, amyotrophic lateral sclerosis) and one metabolic-disease pathway (type-II DM). Thirteen signaling pathways included four pathways involved in the immune system (NOD-like receptor, Toll-like receptor, Fc epsilon RI, B cell receptor), two signal-transduction pathways (ErbB, vascular endothelial growth factor (VEGF)), two cell growth and death pathways (p53 signaling pathway, apoptosis), three endocrine-system pathways (progesterone-mediated oocyte maturation, gonadotropin-releasing hormone (GnRH) signaling pathway, adipocytokine signaling pathway) and one lipid-metabolism pathway (biosynthesis of steroid hormone). The Compound-Pathway network was constructed with 29 candidate compounds and their significant pathways

Table II. Information for candidate active compounds from ASH, RRER and GF herbs.

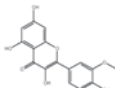
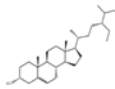
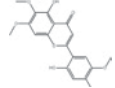
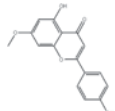
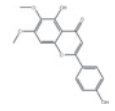
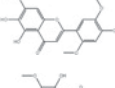
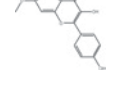
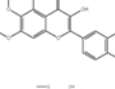
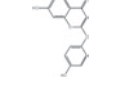
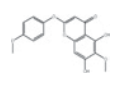
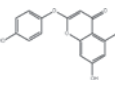
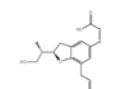
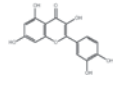
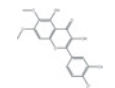
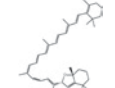
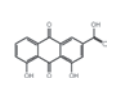
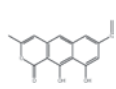








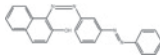
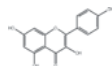






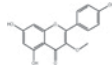
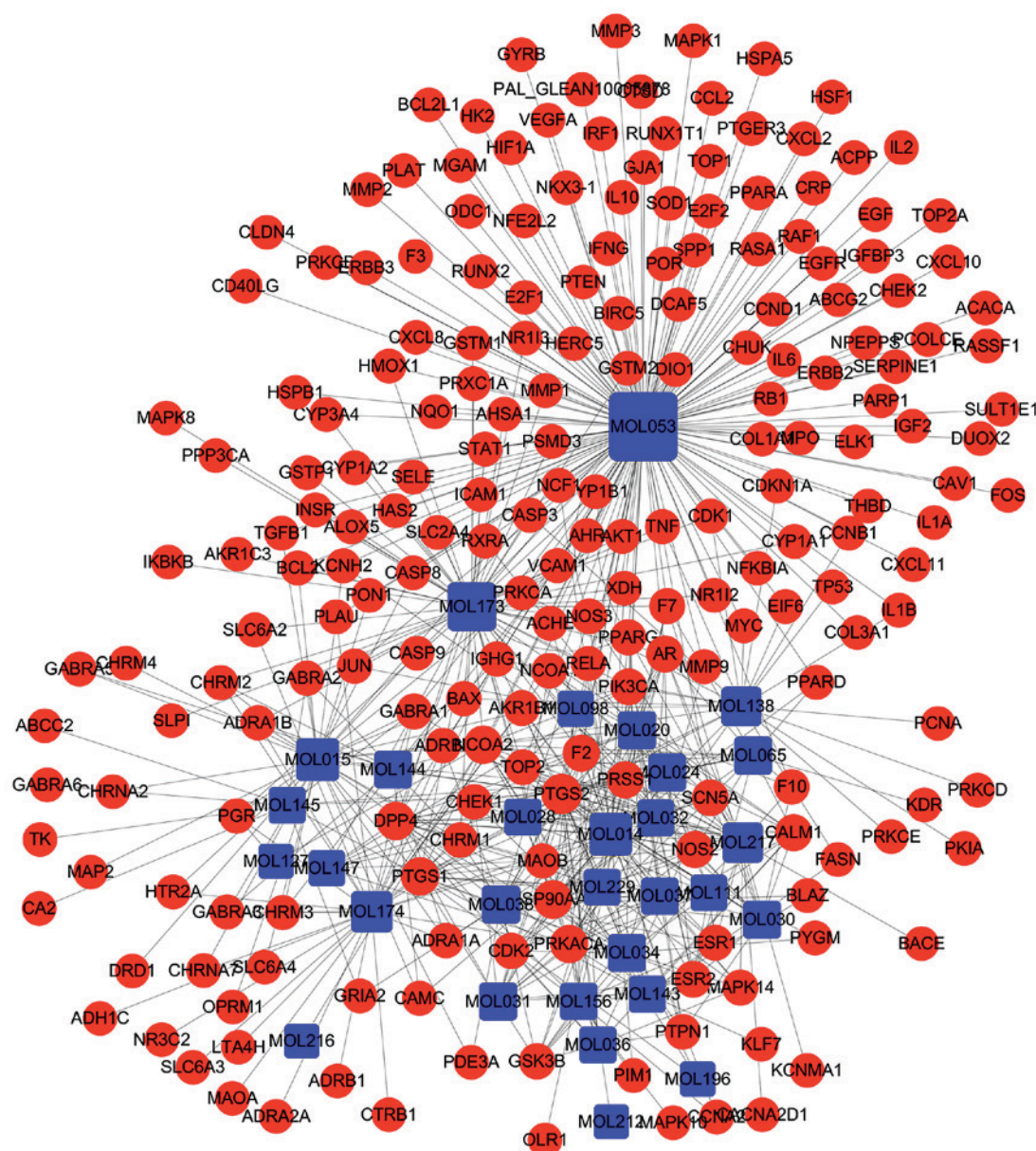
Number	Molecule name	OB (%)	Caco-2	DL	Molecular structure	Herb
MOL014	Isorhamnetin	49.6	0.31	0.31		ASH
MOL015	Beta-sitosterol	36.91	1.32	0.75		ASH/RRER/GF
MOL020	Areapillin	48.96	0.6	0.41		ASH
MOL024	Genkwanin	37.13	0.63	0.24		ASH
MOL028	Skrofullein	30.35	0.72	0.3		ASH
MOL030	Isoarcapillin	57.4	0.4	0.41		ASH
MOL031	Eupalitin	46.11	0.62	0.33		ASH
MOL032	Eupatolitin	42.55	0.16	0.37		ASH
MOL034	Capillarisin	57.56	0.49	0.31		ASH
MOL036	4'-Methylcapillarisin	72.18	0.57	0.35		ASH
MOL037	Demethoxycapillarisin	52.33	0.31	0.25		ASH
MOL038	Artepillin A	68.32	0.45	0.24		ASH
MOL053	Quercetin	46.43	0.05	0.28		ASH/GF
MOL065	Eupatin	50.8	0.53	0.41		RRER
MOL081	Mutatochrome	48.64	1.97	0.61		RRER
MOL098	Rhein	47.07	-0.2	0.28		RRER
MOL111	Toralactone	46.46	0.86	0.24		RRER

Table II. Continued.

Number	Molecule name	OB (%)	Caco-2	DL	Molecular structure	Herb
MOL127	Daucosterol qt	35.89	1.35	0.7		RRER
MOL133	Palmidin A	32.45	-0.36	0.65		RRER
MOL138	Aloe-emodin	83.38	-0.12	0.24		RRER
MOL143	(-)-catechin	49.68	-0.03	0.24		RRER
MOL144	Crocetin	35.3	0.54	0.26		GF
MOL145	Genipin	26.06	-0.37	0.10		GF
MOL146	(4aS,6aR,6aS,6bR,8aR,10R,12aR,14bS)-10-hydroxy-2,2,6a,6b,9,9,12a-heptamethyl-1,3,4,5,6,6a,7,8,8a,10,11,12,13,14b-tetradecahydronicene-4a-carboxylic acid	32.03	0.61	0.76		GF
MOL147	Ammidin	34.55	1.13	0.22		GF
MOL156	Sudan III	84.07	0.42	0.59		GF
MOL173	Kaempferol	41.88	0.26	0.24		GF
MOL174	Stigmasterol	43.83	1.44	0.76		GF
MOL196	Mandenol	42	1.46	0.19		GF
MOL198	Supraene	33.55	2.08	0.42		GF
MOL212	Isoimperatorin	45.46	0.97	0.23		GF
MOL216	Ethyl oleate (NF)	32.4	1.4	0.19		GF
MOL217	5-hydroxy-7-methoxy-2-(3,4,5-trimethoxyphenyl) chromone	51.96	0.88	0.41		GF
MOL229	3-Methylkempferol	60.16	0.37	0.26		GF

OB, oral bioavailability; DL, drug-likeness.



included 60 nodes (29 compounds and 31 pathways) and 468 edges (Fig. 3). Pathways are represented by pink nodes, compounds are represented by blue nodes, and the interactions between them are represented by edges in Fig. 3. Centralization and heterogeneity of the network was 0.273 and 0.533, respectively. It was found that the VEGF signaling pathway (degree=28) was linked with 28 chemical molecules, and that other pathways interacted with at least two molecules.

targets in four diseases: Cancer, metabolic, cardiovascular and immune. The 42 common targets genes were ADH1C, ADRB2, AR, CCL2, COL1A1, CRP, CYP1A1, CYP1A2, CYP3A4, EGF, ESR1, ESR2, F2, GSTM1, GSTP1, HMOX1, ICAM1, IFNG, IGF2, IL10, IL1A, IL1B, IL6, INSR, MMP1, MMP2, MMP9, MPO, NOS2, NOS3, PLA2, PON1, PPARG, PTGS2, SELE, SERPINE1, SLC6A4, SPP1, TGFB1, TNF, TP53, and VEGFA. Among 160 genes, GSTM1, NOS3, and TNF were common genes of 16 diseases.

Discussion

TCM formulations are composed of multiple ingredients. Their mechanism of action is complex, and may be associated

Table III. The 31 significant pathways found by JEPETTO.

Number	Pathway	XD-score	q-value	Overlap/size
1	Bladder cancer	1.47032	0.00000	16/38
2	Non-small cell lung cancer	1.08065	0.00000	16/51
3	Prostate cancer	1.02555	0.00000	26/84
4	Pancreatic cancer	1.01321	0.00000	22/70
5	Endometrial cancer	0.94804	0.00000	14/50
6	Colorectal cancer	0.79509	0.00000	18/61
7	Metabolism of xenobiotics by cytochrome P450	0.77843	0.00045	6/20
8	Glioma	0.76675	0.00000	16/60
9	Leishmaniasis	0.6752	0.00000	17/62
10	Steroid hormone biosynthesis	0.64509	0.00931	4/15
11	Type II diabetes mellitus	0.62494	0.00003	10/43
12	NOD-like receptor signaling pathway	0.61233	0.00000	14/59
13	Prion diseases	0.607	0.00004	9/35
14	Small cell lung cancer	0.57843	0.00000	20/82
15	ErbB signaling pathway	0.56917	0.00000	17/84
16	VEGF signaling pathway	0.56409	0.00000	14/62
17	Chronic myeloid leukemia	0.53594	0.00000	16/69
18	Malaria	0.53081	0.00003	10/42
19	Acute myeloid leukemia	0.52287	0.00003	11/52
20	Chagas disease	0.51782	0.00000	22/99
21	p53 signaling pathway	0.49646	0.00000	15/62
22	Toll-like receptor signaling pathway	0.47843	0.00000	19/90
23	Melanoma	0.46367	0.00003	12/62
24	Apoptosis	0.44093	0.00000	17/81
25	Progesterone-mediated oocyte maturation	0.41387	0.00002	14/79
26	Fc epsilon RI signaling pathway	0.41091	0.00023	11/65
27	B cell receptor signaling pathway	0.40613	0.00002	13/69
28	Graft-versus-host disease	0.37843	0.01012	5/25
29	GnRH signaling pathway	0.37629	0.00003	14/83
30	Adipocytokine signaling pathway	0.3679	0.00036	10/57
31	Amyotrophic lateral sclerosis (ALS)	0.36104	0.00007	10/47

with multiple targets and multiple pathways. YCHD is a classic TCM formulation and used commonly to treat liver diseases by clearing heat, eliminating dampness, and removing jaundice.

In the present study, we proposed a network pharmacologic approach to identify bioactive compounds and significant pathways in YCHD by OB screening, as well as evaluation of DL and intestinal absorption. Finally, 243 compounds in YCHD were extracted from the TCMSP database, and 33 compounds with good OB, DL and small-intestinal absorption were considered to be active molecules in YCHD for further study.

Some of the compounds have been shown to possess pharmacologic activities for the treatment of various diseases. These include the anti-lung-cancer activity of isorhamnetin (25), as well as beta-sitosterol (analgesic) (26), genkwanin (anti-colorectal cancer) (27), eupalitin (anti-prostate carcinoma) (28), capillarisin (anti-hyperalgesic and anti-allodynic) (29), quercetin (anticancer) (30), rhein (anticancer) (31), aloe-emodin (anti-growth disorders) (32), crocetin (anticancer) (33), Sudan III (anti-persistent chylous

ascites) (34), kaempferol (anti-pancreatic cancer) (35), stigmasterol (anticancer) (36). In our study, based on the potential targets that 34 compounds act upon, 31 significant pathways and 16 classes of diseases that were associated with the targets were obtained.

By analyzing the topologic properties of the Compound-Target interaction network, we found that compounds with high degree nodes and protein targets that occupied hub positions in the network could perform important roles in the pharmacologic function of YCHD. Analyses of the Compound-Pathway network showed that the main active ingredients in YCHD could act on multiple pathways, and that the TCM formulations had multiple components, multiple targets and integrated regulation (14). Then, we linked potential genes to diseases, and found that these potential genes were related to several complex diseases: Cancer, cardiovascular, metabolic, and immune. For example, the development and progression of tumors is associated with multiple pathways (37,38). The main advantage of TCM therapeutics is that the many compounds within them exert a more robust

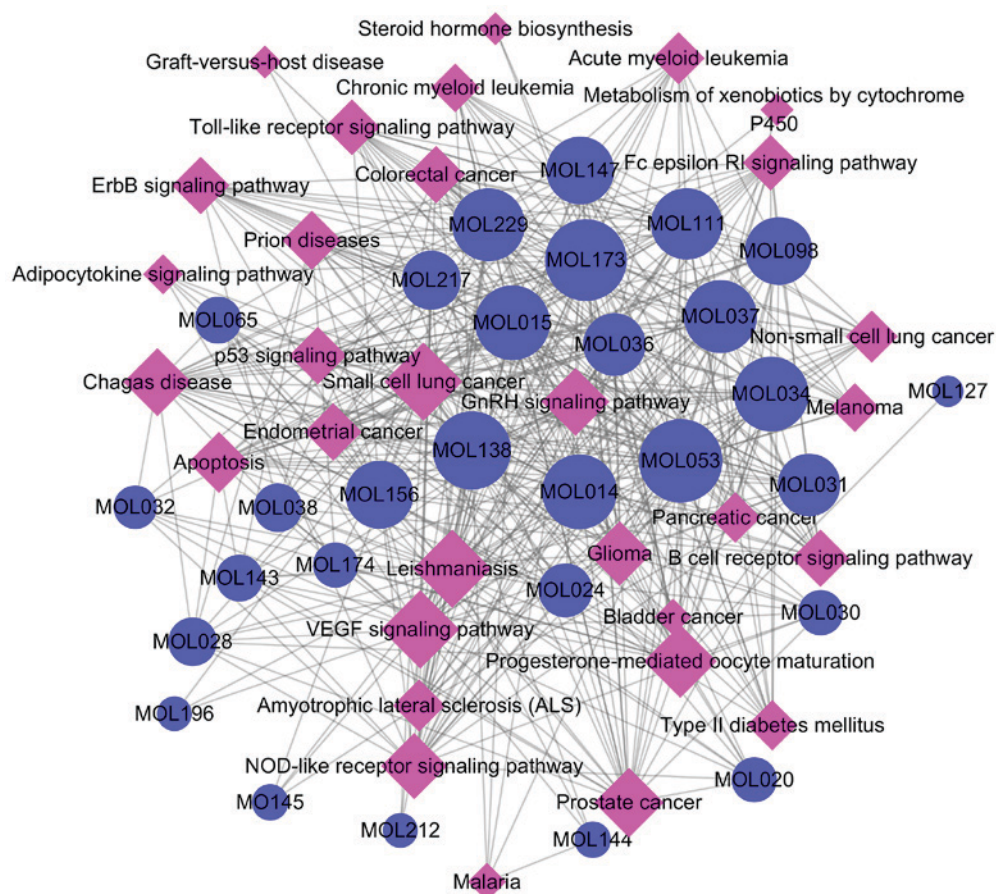


Figure 3. Compound-Pathway network for YCHD. The pink nodes represent significant pathways and the blue nodes represent active compounds. The edges represent the interaction between them and node size is proportional to the degree.

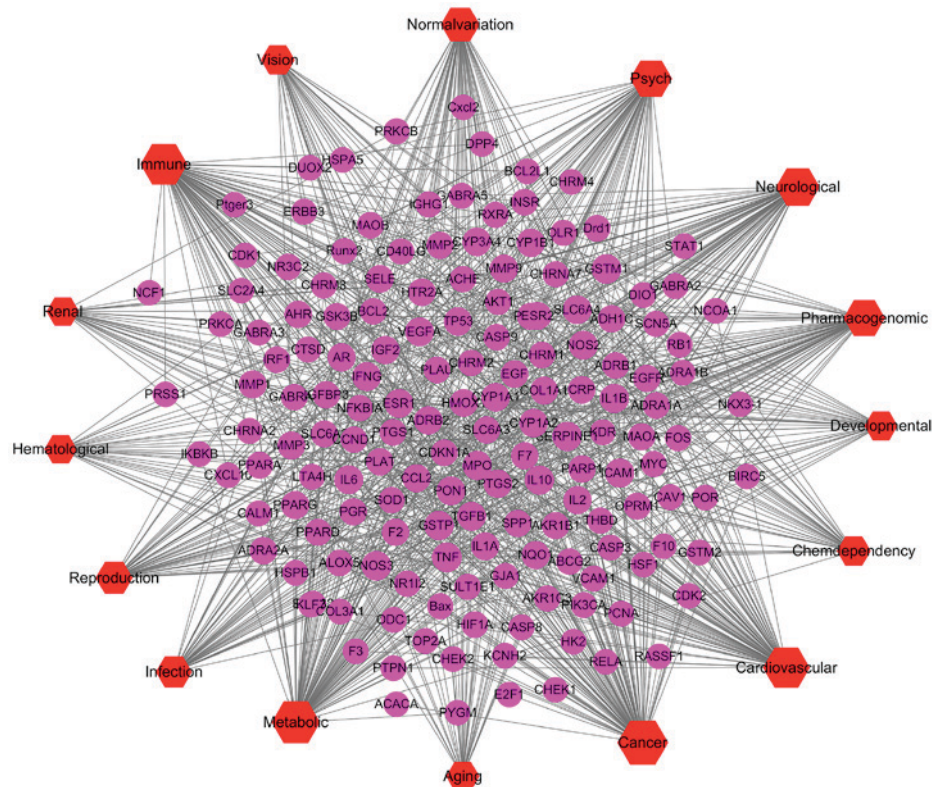


Figure 4. Gene-Disease network for YCHD. The red nodes represent disease and the pink nodes represent genes. The edges represent the interaction between them and node size is proportional to the degree.

synergistic effect than any individual compound by 'hitting' multiple targets. Therefore, the multidirectional mechanisms of TCM formulations that act through multiple pathways of the immune system, endocrine system, signal transduction and cell growth and death may offer a new therapeutic tool to treat tumors.

Studies have shown that the active molecules in YCHD include beta-sitosterol (39), isorhamnetin (25), genkwanin (27), eupalitin (28) and quercetin (30) and have anticancer effects. In particular, quercetin is a potent antioxidant flavonoid found in many common medicinal herbs, and possesses a wide spectrum of biologic activities (40). Quercetin may have cardioprotective, anticancer, anti-ulcer, anti-allergic, anti-viral, anti-inflammatory, anti-DM, gastroprotective, anti-hypertensive, anti-infective and immunomodulatory activities (41). Our network study also showed that YCHD can act on cancer, metabolic, cardiovascular, and immune systems. Disease enrichment analyses showed that 96 genes were related to cancer. Pathway enrichment analyses demonstrated YCHD to be involved with regulation of multiple pathways in cancer, including apoptosis (1) as well as various signaling pathways: NOD-like receptor (42), Toll-like receptor (43), Fc epsilon RI (44), B-cell receptor (45), ErbB (46), VEGF (47), and GnRH (48). Therefore, YCHD may exert anticancer effects through regulation of cell death, anti-inflammation, anti-immune system, anti-angiogenesis, and energy metabolism (37). Our study also verified a report stating that YCHD has pharmacologic activities against primary liver cancer (49), pancreatic carcinoma (1), and DM (9).

In most cases, the occurrence and development of a disease can also be considered to be a result of a network (50,51). Interestingly, we found 42 common targets in four diseases (cancer, metabolic, cardiovascular, and immune) among 160 genes. We also showed that a disease does not occur in isolation. Studies using a one-target and one-drug model tend to ignore the relationship between diseases (52). The constituents of TCM formulations are complex, and the effect that a single component produces may be relatively weak. However, these ingredients with different effects and different targets can act on various aspects of the disease through systems, and they interact with each other to produce synergistic effects (53,54). Network pharmacology can be used to predict the target profiles and pharmacologic actions of herbal compounds. In our study, network-construction approaches were applied to identify bioactive compounds and potential targets and to determine the underlying mechanism of action of YCHD. However, additional experiments must be carried out to validate our study results.

Thirty-four bioactive compounds with 31 significant pathways in YCHD were identified by performing network analyses, which explains how to treat disease through multiple components, targets and pathways. The method of network pharmacology developed in our study could provide novel insights into the mechanism of action of YCHD.

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