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Renshen Yangrong decoction for secondary malaise and fatigue: network pharmacology and Mendelian randomization study

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Background: Renshen Yangrong decoction (RSYRD) has been shown therapeutic effects on secondary malaise and fatigue (SMF). However, to date, its bioactive ingredients and potential targets remain unclear.

Purpose: The purpose of this study is to assess the potential ingredients and targets of RSYRD on SMF through a comprehensive strategy integrating network pharmacology, Mendelian randomization as well as molecular docking verification.

Methods: Search for potential active ingredients and corresponding protein targets of RSYRD on TCMSP and BATMAN-TCM for network pharmacology analysis. Mendelian randomization (MR) was performed to find therapeutic targets for SMF. The eQTLGen Consortium (sample sizes: 31,684) provided data on cis-expression quantitative trait loci (cis-eQTL, exposure). The summary data on SMF (outcome) from genome-wide association studies (GWAS) were gathered from the MRC-IEU Consortium (sample sizes: 463,010). We built a target interaction network between the probable active ingredient targets of RSYRD and the therapeutic targets of SMF. We next used drug prediction and molecular docking to confirm the therapeutic value of the therapeutic targets.

Results: In RSYRD, network pharmacology investigations revealed 193 possible active compounds and 234 associated protein targets. The genetically predicted amounts of 176 proteins were related to SMF risk in the MR analysis. Thirty-seven overlapping targets for RSYRD in treating SMF, among which six (NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1) were prioritized with the most convincing evidence. Finally, the 14 active ingredients of RSYRD were identified as potential drug molecules. The strong affinity between active components and putative protein targets was established by molecular docking.

Conclusion: This study revealed several active components and possible RSYRD protein targets for the therapy of SMF and provided novel insights into the feasibility of using Mendelian randomization for causal inference between Chinese medical formula and disease.

KEYWORDS

Renshen Yangrong decoction, secondary malaise and fatigue, network pharmacology, Mendelian randomization, molecular docking

1 Introduction

Fatigue, in medical terminology, is defined as a notable deficiency in both physical and mental energy that is seen by the subject or caregiver as hindering normal or desirable activities. It can be categorized as either physiological, secondary, or chronic (such as in the case of Myalgic Encephalomyelitis) (1, 2). Secondary malaise and fatigue (SMF) is caused by disease-related symptoms such as depression, decreased activity, sleep disturbance, anemia, hypothyroidism, malnutrition, and infections rather than the disease itself (3, 4). Furthermore, the therapy of the condition, or medication in general, might lead to secondary weariness, particularly when numerous pharmacotherapies ("polypharmacy") are used (5). Currently, treatment mainly targets potential diseases. For example, pulmonary rehabilitation helps the respiratory symptoms of chronic obstructive pulmonary disease and also fatigue (6). Exercise therapy and psychological therapies, notably cognitive behavior therapy, offer some evidence of success in the treatment of fatigue caused by cancer, inflammatory illnesses (e.g., inflammatory bowel disease, rheumatoid arthritis) (7-9), and neurological disorders (for example, multiple sclerosis, myasthenia gravis) (10, 11). Exercise may improve fatigue and function related to fibromyalgia (12, 13). Tai chi may be superior to aerobic exercise for those with fibromyalgia (14). Massage and acupuncture may help manage cancer-related fatigue (15, 16). Understanding SMF involves recognizing it as a symptom rather than a standalone condition. Alongside treating the root cause, patients may benefit from general strategies to reduce fatigue, such as regular physical activity, balanced nutrition, stress management techniques, and pacing daily activities to avoid overexertion (17). Effectively managing SMF requires a comprehensive, individualized approach that targets the specific underlying cause while also incorporating general strategies to improve overall energy levels and well-being. There is no evidence that pharmacologic treatment targeting fatigue (e.g., modafinil, methylphenidate) helps manage fatigue related to most chronic diseases (18, 19). Thus, further efforts are still urgently needed to develop novel strategies to prevent this secondary malaise and fatigue.

Zhubing Yuanhou Lun, a well-known tract about the etiology and symptoms of disease produced during the Sui Dynasty, contains almost 70 different types of symptoms in its chapter on "consumptive disease." The symptoms can be divided into two categories: psychological symptoms like depression, anxiety, restlessness, and so on, and somatic symptoms like fatigue, a somatic sense of heaviness, and somatic pain (joint and muscle pain), among others. Traditional Chinese Medicine (TCM) categorizes SMF as a "consumptive disease." Patients may have poor blood and Qi due to extended exposure to pollutants, overwork, emotional distress, or incorrect food. This can lead to renal function impairment and five-organ failure, ultimately culminating in SMF. Additionally, blood and Qi impairment are aggravated by chemotherapy, radiotherapy, and surgery, which exacerbates fatigue symptoms. Thus, Qi/blood deficiency and dysfunction of the five organs are implicated within the aetiology of SMF in TCM. To make up for the shortage, SMF is primarily treated by tonifying the body and strengthening its resistance. RSYRD (Ninjin'yoeito in Japanese) was first documented in Taiping Huimin Heji Jufang, a classic ancient Chinese medical text first published in 1078 during the Song Dynasty. Its efficacy in treating anemia (20), fatigue (21-27), general malaise brought on by chemotherapy and malignant tumours (25, 27), as well as psychological conditions like anxiety and depression and the lethargic and apathetic symptoms of Parkinson's and Alzheimer's diseases (28-31), has been demonstrated in earlier clinical and preclinical research. Despite extensive research, due to the diverse composition and functions of Chinese herbal medicines, the specific mechanisms by which they exert their corresponding effects are difficult to explore through basic experiments, limiting the widespread clinical application of RSYRD worldwide.

Incorporating genetics into the production of medications is a highly effective approach to expedite this process, as medicines based on genetics have a far higher likelihood of success in clinical trials (32, 33). Druggable genes that encode proteins have now been identified as possible targets for pharmaceuticals, small compounds, and monoclonal antibodies (34, 35). GWAS can effectively identify single nucleotide polymorphisms (SNPs) related with disease risk, but it cannot reliably identify causal genes and drive therapeutic development (36, 37). The integration of network pharmacology and Mendelian randomization provides a practical approach to exploring the pharmacology mechanism of Chinese medical formulae and evaluating the pharmacological modulation of the gene target. Drug target Mendelian randomization is an application of MR in drug target validation. It employs genetic instruments linked to drug target genes to determine whether there is a link between the medication target and disease outcome (34, 38). Network pharmacology can systematically reveal the active ingredients in Chinese herbal medicines and predict the relationships between drug compounds and protein targets (39). Molecular docking could be used to validate the interaction between active components and key therapeutic targets (40). Finally, our findings provide vital insights into new theoretical support for the therapeutic therapy of SMF. We provide informative advice for the development of more effective and focused therapeutic modalities by integrating drug target Mendelian randomization, network pharmacology, molecular docking, and drug prediction.

2 Materials and methods

2.1 Composition of Renshen Yangrong decoction

RSYRD is a decoction of 12 botanicals after water extraction, which includes Renshen (*Panax Ginseng C. A. Mey.*), Baizhu (*Atractylodes Macrocephala Koidz.*), Chenpi (*Citrus Reticulata*), Gancao (*licorice*), Fuling (*Poria Cocos (Schw.*) Wolf.), Huangqi (*Hedysarum Multijugum Maxim.*), Baishao (*Paeoniae Radix Alba*), Danggui (*Angelicae Sinensis Radix*), Shudihuang (*Rehmanniae Radix*

Abbreviations: GAA, Alpha glucosidase; cis-eQTL, Cis-expression quantitative trait loci; DL, Drug-like characteristics; GWAS, Genome-wide association studies; IVW, Inverse variance weighted; MMP9, Matrix metalloproteinase-9; KCNH2, Potassium voltage-gated channel subfamily H member 2; OB, Oral bioavailability; PDB, Protein Database; RSYRD, Renshen Yangrong decoction; MR, Mendelian randomization; SIRT1, Sirtuin 1; SLC16A1, Solute carrier family 16 member 1; SMF, Secondary malaise and fatigue; SNPs, Single nucleotide polymorphisms; NO, Nitric oxide; NOS, Nitric oxide synthase; NOS3, Nitric oxide synthase 3; RB1, Retinoblastoma susceptibility gene; TCM, Traditional Chinese Medicine; TCMSP, Traditional Chinese Medicine Systems Pharmacology; VNTRs, Variable number of tandem repeats.

Praeparata), Wuweizi (*Schisandrae Chinensis Fructus*), Rougui (*Cinnanmomi Cortex*), Yuanzhi (*Polygala tenuifolia Willd*.), and they all possess medicinal values. The drug names refer to the 2020 edition of the Chinese Pharmacopeia and the plant classification refers to the MPNS database.¹

2.2 Screening the active ingredients and targets of RSYRD

The active components of RSYRD, as well as their related targets, were collected from the Traditional Chinese Medicine Systems Pharmacology (TCMSP) Database and Analysis Platform² by using "Renshen," "Baizhu," "Fuling," "Gancao," "Chenpi," "Huangqi," "Danggui," "Baishao," "Shudihuang" and "Wuweizi" as keywords. However, the chemical components of "Yuanzhi" and "Rougui" were taken through the BATMAN-TCM database³ (41) and the obtained compounds were then further screened by TCSMP. The TCMSP database advised screening bioactive compounds based on threshold values of drug-like characteristics (DL) \geq 0.18 and oral bioavailability (OB) \geq 30% (42). The UniProt database⁴ was used to standardize all target names.

2.3 Exposure and outcome dataset

The eQTLGen consortium⁵ provided the whole set of cis-eQTLs data and allele frequency statistics (43). The 4,463 genes on the list of druggable genes are from a previous study (35, 44). The eQTLs analyzed in our study were restricted to SNPs positioned within a 100 kilobase (kb) range before or after the endpoint of a druggable gene. This selection criterion was based on the proximity to the gene of interest in drug development research and the potential for more direct regulation of gene expression. At last, eQTLs for 1,450 druggable genes have been identified. We used GWAS analysis of SMF from MRC-IEU Consortium as outcome variables, including genotype data of 1,245 SMF patients and 461,765 controls.⁶

2.4 Mendelian randomisation analysis

MRPRESSO and TwoSampleMR (version 0.5.6) were the R packages used to perform MR studies (45). Initially, we aligned the instrumental variables that target exposure-related drugs with the outcome datasets. Subsequently, we conducted analysis using several methods including MR Egger, weighted median, inverse variance weighted (IVW), simple mode, weighted mode, and MR-PRESSO. Among these approaches, the IVW method was the most commonly employed. The MR Egger and IVW techniques were used to perform the heterogeneity test (46). The Q value of Cochrane

was used to assess the heterogeneity of genetic tools, and p > 0.05 indicated that there was no substantial heterogeneity. The MR Egger regression equation was used to assess the genetic tool's horizontal pleiotropy, and p > 0.05 indicated that there was no horizontal pleiotropy (47). After removing the outlier with the MR-PRESSO test, the sensitivity analysis was repeated. To ensure that no SNP had a substantial impact on our results, we utilized the leave-one-out approach to delete each SNP in turn and compared the IVW method results with all variants. FDR-corrected p values were calculated, and FDR of <0.05 was considered significant.

2.5 Herb-ingredient-target network construction

A network of interactions between herbs, ingredients, and targets was generated by inputting active substances in RSYRD, common targets of these active compounds, and SMF into the Cytoscape 3.9.1 program (48). Node represents active ingredients, herbs, and targets, while edge represents the relationship between different nodes. The potential core active ingredients of RSYRD in the therapy of SMF were conjectured using the Cytoscape software's Network Analyzer tool in conjunction with the primary active ingredients, core targets, and key herbs. Furthermore, the multi-component and multitarget synergistic effects of Chinese herbal medications were assessed in the context of SMF-related networks.

2.6 GO and KEGG pathway enrichment analyses

To investigate the role of identified prospective the rapeutic target genes in functional characteristics and biological mechanisms, we conducted Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway enrichment analysis using the "cluster Profiler" (49) and "Pathview" (50) package of R software (version 4.1.3). GO includes three terms: biological process (BP), cellular component (CC), and molecular function (MF), displaying the top 10 terms of each category, respectively. The KEGG pathway can provide metabolic pathway information. The results of all biological pathway enrichment analyses were visualized using the "ggplot2" package. The *p*<0.05 of enriched terms were considered significantly enriched.

2.7 Protein–protein interaction and druggability prediction

A PPI network was constructed using the STRING database⁷ to investigate potential interactions among the identified proteins. The discovered proteins have been further analysed for possible use as therapeutic targets by searching Drug Signatures Database⁸ for druggene interactions (51). This database prioritised druggable targets

7 https://string-db.org/

8 http://dsigdb.tanlab.org/DSigDBv1.0/

¹ http://mpns.kew.org

² http://tcmspw.com/tcmsp.php

³ http://bionet.ncpsb.org/batman-tcm/

⁴ https://sparql.uniprot.org/

⁵ https://eqtlgen.org/

⁶ https://gwas.mrcieu.ac.uk/datasets/ukb-b-11972/

through the integration of information gathered through text mining, expert curation, drug-gene interactions, as well as gene function. The Enrichr platform⁹ analyzed and visualized data (52).

2.8 Molecular docking of active ingredients with core targets

The 2D structures of all the substances were obtained from the PubChem database¹⁰ and stored in "SDF" format. The software Chem 3D was utilized to transform the "SDF" file format into mol2 structures, specifically for small molecule ligands. The key targets' 3D structures were retrieved from the Protein Database (PDB)¹¹ and saved as protein receptors in "PDB" format. The PyMOL program (version 2.4.0) was used to isolate the original ligands from the main target proteins. The AutoDock software (version 4.2.0) was used for importing the processed protein targets and performing hydrogenation, total charge estimation, as well as atom type setting. Protein receptors and ligands were recorded using the PDBQT format. AutoDock-Vina software (version 1.5.6) was used for molecular docking in order to provide a comprehensive score of the affinity of the receptor-ligand complexes. Heatmaps were plotted to visualize these values in R. The PyMOL software served for visualizing the docking results.

2.9 Identification of DEGs and clinical characteristics of key targets

To obtain clinical implications of potential key targets, we searched gene expression changes in Peripheral Blood Mononuclear cells (PBMC) induced by physical activity in the GEO database.¹² The GEO dataset GSE12385 was used to validate the key genes. With |log2 fold change (FC)| >1 and p<0.05 as screening criteria, differentially expressed genes (DEGs) from GSE12385 were identified utilizing "Limma" R package, where log FC >1, p<0.05 was Up, log FC <-1, p<0.05 was Down. The heatmap, volcano plot and box plot were conducted using "heatmap" and "ggplot2" packages of R software. Flow chart of this study was presented in Figure 1.

3 Results

3.1 Active ingredients in RSYRD and targets screening

After screening, a total of 193 active ingredients in RSYRD from TCMSP and BATMAN-TCM databases were found: 22 from Renshen, 7 from Baizhu, 15 from Fuling, 92 from Gancao, 5 from Chenpi, 20 from Huangqi, 2 from Danggui, 13 from Baishao, 2 from Shudihuang, 8 from Wuweizi, 3 from Rougui, and 4 from Yuanzhi. After normalizing the protein targets from the TCMSP database using the Uniprot database, 834 gene targets have been identified, containing 101 from Renshen, 18 from Baizhu, 19 from Fuling, 217 from Gancao, 62 from Chenpi, 196 from Huangqi, 48 from Danggui, 85 from Baishao, 28 from Shudihuang, 18 from Wuweizi, 26 from Rougui, and 16 from Yuanzhi. Following the merger, 234 distinct values remained after 600 redundant gene targets were eliminated.

3.2 Candidate druggable genes for SMF

The study examined the MR correlations between 1,450 proteins that have accessible index eQTL signals and the risk of SMF outcomes. The results of this investigation may be seen in the Supplementary material. We found 176 protein-SMF pairings with marginal significance (p < 0.05). After removing and integrating false positive and repetition targets, 48 prospective RSYRD targets against SMF were found. After MR quality control it was shown that 37 candidate druggable genes have the most robust MR evidence for SMF risk (Figure 2). Figure 3 depicts an overview of the MR analysis results, which identify 17 prospective druggable genes for SMF after FDR adjustment (FDR <0.05) (Supplementary Table S1). Ultimately, NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1, as potential targets of RSYRD for SMF reached FDR <0.05 (Figure 4A). Genetic predictions indicate that elevated levels of IMPA1, P4HTM, and NOS3 are linked to a higher risk of SMF. Conversely, lower levels of GAA, RB1, and SLC16A1 are associated with an increased risk of SMF. This suggests that the three proteins are negatively correlated with SMF risk (Figure 4B). In addition, the sensitivity analysis indicated that there was no variation or cross-effects in any of the other outcomes (p > 0.05). The leave-one-out technique revealed that eliminating any SNP for SMF made no change in the results (Figure 5).

3.3 H-I-T network construction and topological network analysis

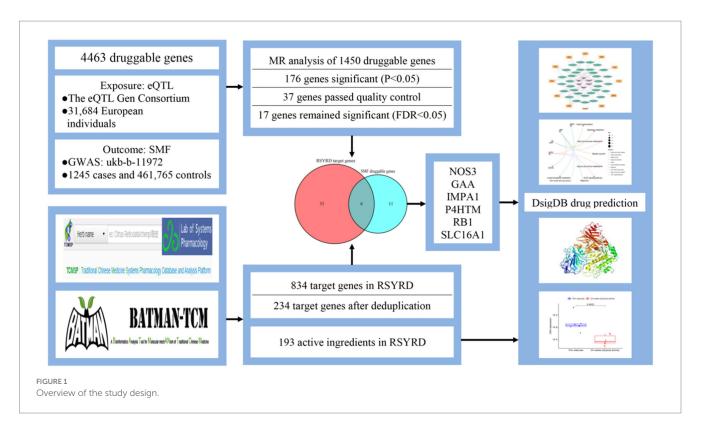
On a system level, the H-I-T network with 135 nodes and 282 edges was built by Cytoscape 3.9.1 (Figure 6) to gain a better knowledge of the interaction between 12 herbs, 86 ingredients, and 37 intersection target genes. To assess node degree values, we used Cytoscape's built-in NetworkAnalyzer. Among these, the active components in Gancao with the greatest number of associated targets were Quercetin (degree = 14), Kaempferol (degree = 8), and Naringenin (degree = 6), and the active ingredients with more targets in Huangqi, Renshen, and Danggui were Curcumin (degree = 13), Resveratrol (degree=9), Capsaicin (degree=7), Acetaldehyde (degree=6), BETA-ELEMENE (degree=6) and Oleic Acid (degree=6). Furthermore, the targets connecting more active ingredients included matrix metalloproteinase-9 (MMP9) (degree = 25), potassium voltage-gated channel subfamily H member 2 (KCNH2) (degree=21), nitric oxide synthase 3 (NOS3) (degree = 17), alpha glucosidase (GAA) (degree = 12) and sirtuin 1 (SIRT1) (degree = 10). Further analysis shows that the herbingredient-target network comprised a total of 50 nodes (12 herb

⁹ https://amp.pharm.mssm.edu/Enrichr/

¹⁰ https://pubchem.ncbi.nlm.nih.gov/

¹¹ https://www.rcsb.org/

¹² http://www.ncbi.nlm.nih.gov/geo/



nodes, 32 ingredient nodes, 6 target nodes) and 91 edges (Figure 7). Huangqi (degree = 10) and Renshen (degree = 7) are the main effective ingredients. Table 1 lists the effective components and indications of crude herbs in RSYRD for the treatment of SMF (53–61, 63–66).

3.4 GO and KEGG pathway enrichment analyses

We conducted biological pathway enrichment analysis of 17 druggable genes obtained through Mendelian Randomization to explore the prospective therapeutic target genes of SMF, and further analyzed NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1 to clarify the potential mechanism of RSYRD in treating SMF. As shown in Figure 8A, the most significant pathways in the BP category were all associated with regulation of the force of heart contraction, oligosaccharide metabolic process, lipopolysaccharide-mediated signaling pathway, homeostasis of number of cells, and glutamine family amino acid metabolic process. In class CC, drug target genes are similarly enriched for endoplasmic reticulum-related components (immunological synapse, SWI/SNF complex, endoplasmic reticulum quality control compartment, and endosome lumen), which is consistent with previous studies (67). Furthermore, in terms of MF, these genes are also involved in functions strongly associated with energy metabolism (hydrolase activity, mannosyl-oligosaccharide mannosidase activity, and oxidoreductase activity, acting on NAD (P) H, heme protein as acceptor). As shown in Figure 8B, the four pathways analysed by KEGG enrichment are insulin resistance, arginine biosynthesis, diabetic cardiomyopathy and galactose metabolism, all four of which are closely linked to the metabolic syndrome. Similarly, NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1 are associated with metabolic pathways such as arginine and proline metabolism, galactose metabolism, starch and sucrose metabolism, and inositol phosphate metabolism (Figure 8C) (68).

3.5 PPI and druggability prediction on therapeutic target's potential

The protein–protein interaction analysis discovered just one interaction between the identified potential causative proteins, P4HTM and GAA, which were engaged in signal transduction and glucose metabolism pathways, respectively. The present research utilized the DSigDB database to predict potentially useful interventional drugs. The following proteins (NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1) were targeted for drug development, according to our evaluation of druggability. We used Enrichr to identify possible drug molecules based on transcriptome characterization in the DSigDB database. After taking the intersection with the active ingredients of RSYRD, 14 potential drugs were recommended to act on target genes based on adjusted *p*-value (Table 2). This indicates that the active ingredients of RSYRD can be used for the treatment of SMF.

3.6 Validation of the relationship between RSYRD ingredients and core targets by molecular docking

Further molecular docking simulations were performed to further investigate the interactions between the active ingredients of RSYRD and core targets (NOS3, GAA, RB1, and SLC16A1) and to elucidate the Traditional Chinese medicine formulas development plan. According to the molecular docking data, the binding energy among the core target

Gene	nsnp	pval		OR(95% CI)
ABCC1	7	0.029	⊢ ••	1.0007 (1.0001 to 1.0013)
ABCC3	11	0.009	•	0.9998 (0.9996 to 0.9999)
ABCC5	8	0.046	•	1.0002 (1.0000 to 1.0004)
ADAM17	6	0.047	H e t	0.9995 (0.9991 to 1.0000)
AGER	6	<0.001	⊢ ●	1.0015 (1.0006 to 1.0024)
CACNB2	6	0.048	⊢ •-1	1.0008 (1.0000 to 1.0015)
CAMK4	4	0.010	H - -1	0.9991 (0.9984 to 0.9998)
CCNE2	4	0.029	⊢ − −i	0.9988 (0.9978 to 0.9999)
CD36	12	0.027		1.0003 (1.0000 to 1.0006)
CPQ	10	0.023	H	0.9994 (0.9989 to 0.9999)
CPT1A	5	0.041	H	0.9995 (0.9990 to 1.0000)
F2RL1	7	0.021)	1.0003 (1.0001 to 1.0006)
FCER1A	4	0.018	⊢ ••	1.0006 (1.0001 to 1.0012)
GAA	11	<0.001	•	0.9997 (0.9996 to 0.9999)
GRIN3A	5	0.005	H -	1.0010 (1.0003 to 1.0017)
GSTM1	9	0.010	•	0.9997 (0.9995 to 0.9999)
GSTM2	9	0.004	•	0.9996 (0.9994 to 0.9999)
IL10RA	4	0.047	i-●i	1.0009 (1.0000 to 1.0018)
IL15	5	0.020		1.0011 (1.0002 to 1.0020)
IMPA1	5	<0.001	ю	1.0006 (1.0003 to 1.0009)
KCNH2	7	0.012		0.9990 (0.9983 to 0.9998)
LY96	4	0.023	H O H	0.9994 (0.9989 to 0.9999)
MAPK9	16	0.024	•	1.0004 (1.0001 to 1.0007)
MMP1	8	0.021	⊢ ● ⊣i	0.9991 (0.9983 to 0.9999)
MMP9	4	0.034		1.0010 (1.0001 to 1.0019)
NOS3	5	<0.001	⊢	1.0014 (1.0006 to 1.0023)
P2RX7	7	0.004	H	1.0007 (1.0002 to 1.0011)
P4HTM	4	<0.001	H	1.0007 (1.0004 to 1.0010)
PPP5C	13	0.008	•	1.0003 (1.0001 to 1.0006)
PTGDS	7	0.044	i e i	1.0005 (1.0000 to 1.0009)
RB1	4	<0.001	H - H	0.9988 (0.9982 to 0.9995)
SIRT1	4	0.048	•	1.0002 (1.0000 to 1.0004)
SLC16A1	6	<0.001	H -	0.9987 (0.9980 to 0.9994)
SLC22A4	5	0.039	•	0.9998 (0.9995 to 1.0000)
SLC29A1	7	0.028	⊢ ⊕–-i	0.9990 (0.9980 to 0.9999)
SOAT1	5	0.036	-	1.0003 (1.0000 to 1.0006)
TG	5	0.015	H e H	0.9994 (0.9990 to 0.9999)

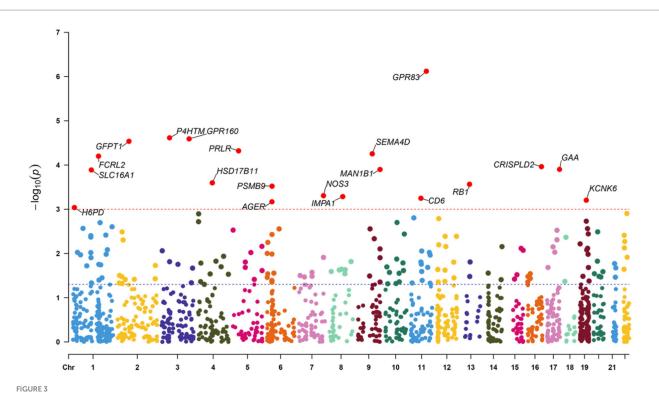
FIGURE 2

The forest plot displays the results of 37 candidate druggable genes.

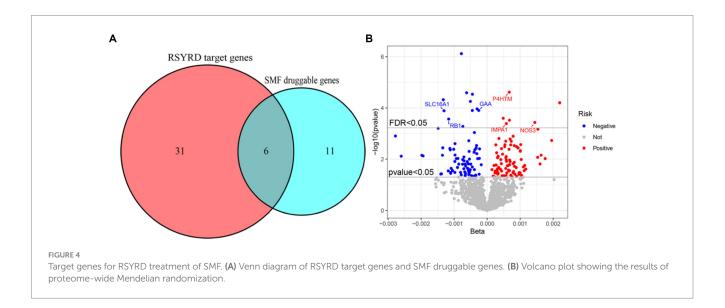
proteins and the effective chemically active ingredients was below -5 kcal/mol, indicating that the binding activity among the core target proteins and the functional core ingredients was stable (Figure 9). Figure 10 shows the docking results of the active ingredients with the best binding affinity for each core target protein. Taken together, these RSYRD representative compounds may bind well to the aforementioned core SMF targets, all of which may play important roles in SMF treatment.

3.7 Identification of DEGs and clinical characteristics of key targets

To evaluate the intra-group data repeatability, we performed principal component analysis (PCA), and the results showed that the repeatability of data in GSE12385 is fine (Figure 11A). Subsequently, gene expression changes in Peripheral Blood Mononuclear cells



Manhattan plot of Mendelian randomization analysis. The blue line represents the nominal significant threshold of 0.05. The red line represents the false discovery rate threshold of 0.05.

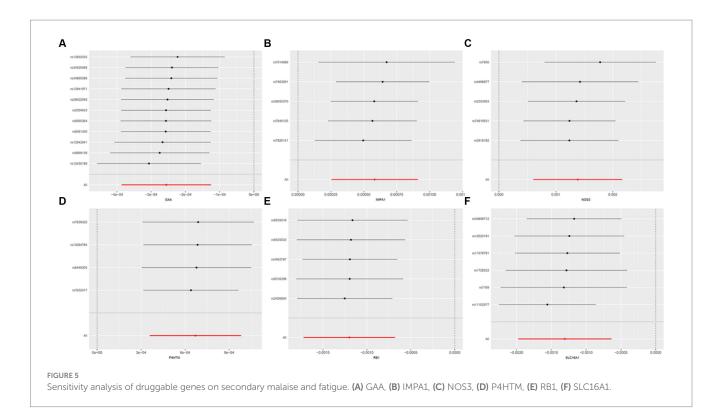


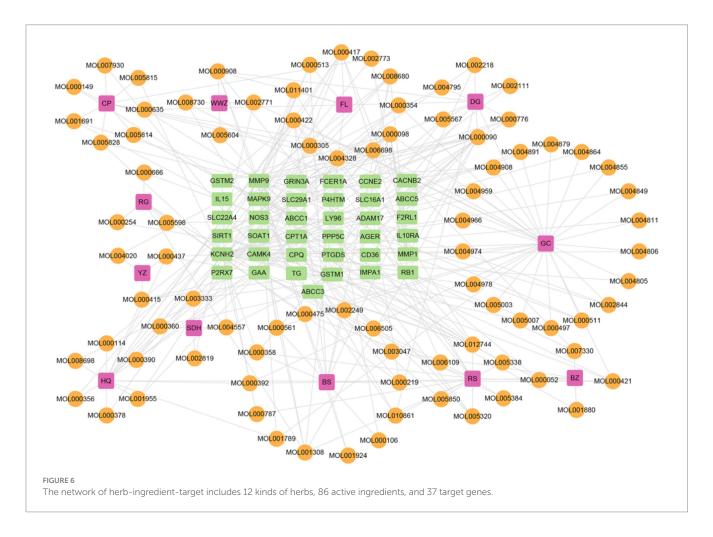
(PBMC) induced by physical activity was investigated in sedentary middle-aged men who undertook a 24-weeks physical activity programme, and to evaluate the expression changes of 6 key targets in blood sampling. Following the analysis of the GSE12385 dataset with R software, the differentially expressed genes between the pre-exercise period and at the end of 24-weeks prescribed physical activity were presented in volcano plot (Figure 11B). Heatmap displays the expression of six key genes (Figure 11C). Moreover, box plots showed the expression patterns of 3 key genes (Figures 11D–F). The results showed a significant increase in mRNA expression of GAA and SLC16A1 in blood sampling among baseline males. On the contrary,

P4HTM was significantly upregulated at the end of the prescribed 24 weeks physical activity.

4 Discussion

In order to explore and understand the therapeutic mechanisms of Traditional Chinese Medicine, we establish a framework that can connect classical Chinese medicine theory to modern biomedical science and can handle the complexity of herbs' chemical composition and therapeutic protein target. Additionally, this is the initial research





to use Mendelian randomization in the study along with network pharmacology. First, network pharmacology was used to examine the active components and protein targets of RSYRD for SMF therapy. We can analyze the "herb-ingredient-proteins/genes-disease" interacting network through the perspective of systems biology using

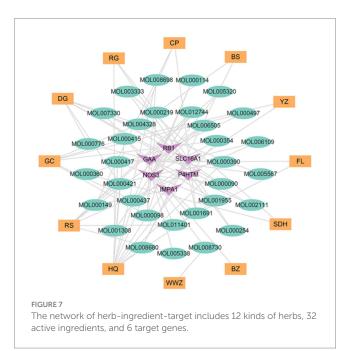
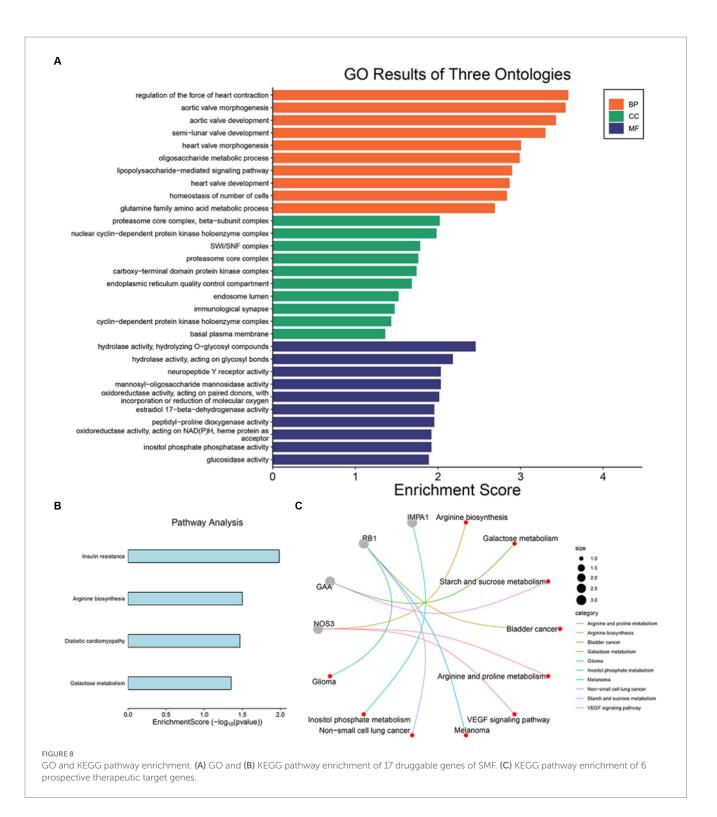


TABLE 1 Effective components and indications of crude herbs in RSYRD.

biological databases as well as clinical trial findings, which will offer knowledge of how the actions of herbs/TCM recipes treat disease and ZHENG. Furthermore, we employ the Mendelian randomization technique along with extensive genome-wide association research data to identify novel therapeutic targets (NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1) for SMF. This approach enhances the efficacy of treatment prediction by demonstrating a cause-and-effect relationship between exposure and result by the utilization of cis-eQTL and SMF risk association data. Finally, to demonstrate the druggability of these target genes, we predicted and molecular docked the medicines corresponding to these targets.

According to the results achieved here, four protein targets in SMF and 14 active ingredients in RSYRD were identified, suggesting that RSYRD exerted its pharmacological effects on treating SMF through multi-ingredients and multi-targets. RSYRD is a formula composed of 12 different herbs, with Huangqi and Renshen forming its most effective ingredients. In traditional Chinese medicine, Huangqi is frequently used for treating patients who have Qi deficiency as well as has been demonstrated to have a strong antifatigue effect (69-71). Renshen was a potential treatment for exhaustion in chronic illness, with few documented side effects (72). Nonetheless, there is little data to support the effectiveness of ginseng supplements in reducing fatigue and boosting athletic performance (73-76). Previous studies have identified polysaccharides, polyphenols, flavonoids, terpenes, peptides, and other components extracted from Chinese herbal medicine as potential anti-fatigue agents (77), quercetin (78), curcumin (79),

Crude herb	Crude herb Effective component Indication of crude herb						
	Effective component						
Renshen	Ginsenoside Re, Chikusetsusaponin IVc, Arachidonic	Anti-aging, anti-diabetic, immunoregulatory, anti-cancer, neuroregulation, wound and					
	Acid, Nicotinic Acid, Pyruvic Acid, Resveratrol,	ulcer healing activity (53)					
	(–)-Epicatechin						
Baizhu	L-Menthol	Immunomodulatory, antitumour, gastroprotective and intestinal health-promoting,					
		hepatoprotective, hypoglycaemic as well as other activities (54)					
Chenpi	L-Ascorbic Acid, Naringenin, Inositol	Antioxidant, anti-inflammatory, antibacterial properties, and anti-cancer activity,					
		hypolipidemic, antiplatelet activity (55, 56)					
Gancao	Licochalcone A, Isoquercitrin, Rutin, Isorhamnetin,	Adrenocortical hormone-like effects, anti-inflammatory, anti-cancer and					
	Naringenin, Quercetin	immunomodulatory effects (57)					
Fuling	Capsaicin, Ginsenoside Re	Anti-tumor, anti-bacterial, anti-oxidant, anti-inflammatory, immunomodulation, and liver					
		and kidney protection (58)					
Huangqi	Chlorogenic Acid, Daidzein, Dihydrocapsaicin,	Antioxidant, anti-inflammatory, immunoregulatory, anticancer, hypolipidemic,					
	Nicotinic Acid, Acetaldehyde, Capsaicin, Quercetin,	antihyperglycemic, hepatoprotective, expectorant, and diuretic effects (59)					
	Isoquercitrin, Isorhamnetin, Rutin						
Baishao	Acetaldehyde, Benzoic acid	Anti-inflammatory and immunomodulatory effects, stimulate blood circulation and					
		exhibit, antiplatelet, and vasodilator activities (60)					
Danggui	Nicotinic Acid, Naringenin, Ethanol, Curcumin, (Z)-	Inhibition of inflammatory factor release, anti-oxidative injury, and interference with					
	Butylidenephthalide, Retinol	collagen production (61)					
Shudihuang	Ferulic acid, Acteoside	Analgesia, sedation, anti-tumor, anti-inflammation, antioxidation, immunomodulation,					
		cardiovascular and cerebrovascular regulation, and nerve damage repair (62)					
Wuweizi	Nordihydroguaiaretic acid	Antioxidants, anti-inflammatory, antiviral, anticancer, and anti-aging effects (63)					
Rougui	Benzoic acid, Oleic acid, Eugenol	Antioxidant, antitumor, analgesics, antihypertension, anticoagulant, anti-inflammatory,					
		antidiabetic effects (64)					
Yuanzhi	Isoquercitrin, Rutin	Antioxidation, anti-inflammation, antidementia, and anti-aging (65)					

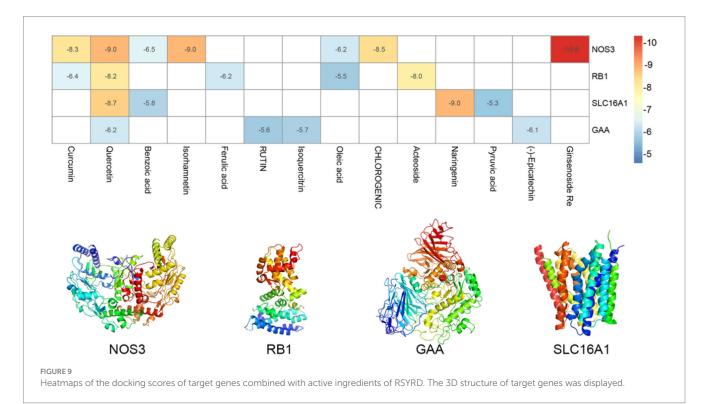


ginsenoside (80), rutin (81), chlorogenic acid, ferulic acid, isoquercitrin, and other natural products have been shown to exhibit definite anti-fatigue properties (77). Nevertheless, Chinese medical formulae are not a patchwork of drugs with the same properties, and there are strict principles for making prescriptions. The reasons for this are that the action of a single drug is usually limited, and some of them might cause adverse effects or even toxicity. However, when numerous medications are used together, ensuring that their

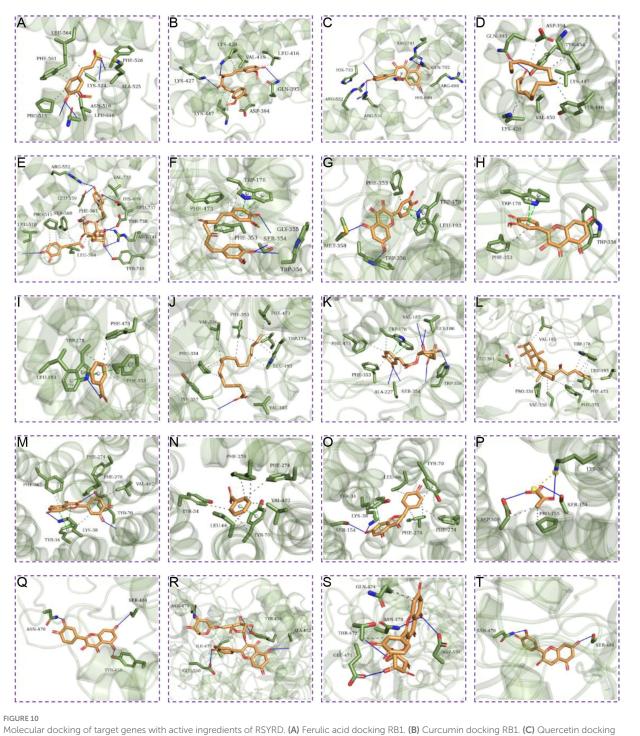
advantages are fully utilized while limiting their disadvantages, they will demonstrate their superiority over a single drug in the treatment of diseases (82). Single-target intervention has been shown to be ineffective and insufficient in complex diseases with robust biological networks, such as cancer (39). In these circumstances, Chinese medicinal formulas techniques can simultaneously target numerous disease targets, resulting in chemical-protein interaction (83, 84).

TABLE 2 Candidate drug predicted using DSigDB.

Term	<i>p</i> -value	Adjusted <i>p</i> -value	Genes
Benzoic acid CTD 00007316	0.001	0.003	SLC16A1; NOS3
Oleic acid CTD 00007269	0.000	0.005	RB1; NOS3
Pyruvic acid CTD 00007265	0.003	0.027	SLC16A1
Rutin TTD 00010730	0.004	0.027	GAA
Isoquercitrin TTD 00008703	0.004	0.027	GAA
(-)-Epicatechin TTD 00000019	0.004	0.027	GAA
Ginsenoside Re CTD 00002168	0.005	0.027	NOS3
Isorhamnetin CTD 00002092	0.005	0.027	NOS3
Ferulic acid CTD 00000186	0.005	0.027	RB1
Quercetin CTD 00006679	0.007	0.027	RB1; SLC16A1; NOS3; GAA
Curcumin CTD 00000663	0.010	0.030	RB1; NOS3
Chlorogenic acid CTD 00005640	0.012	0.030	NOS3
Naringenin CTD 00000211	0.013	0.030	SLC16A1
Acteoside CTD 00002463	0.013	0.030	RB1

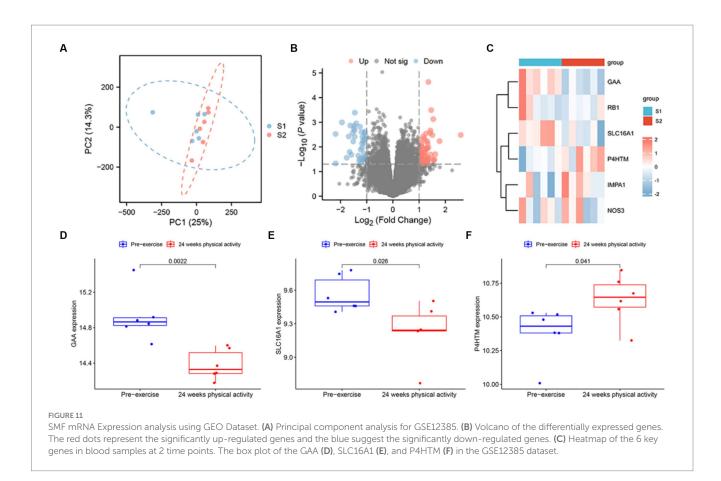


In this study, SLC16A1, GAA, NOS3, and RB1 were determined to be the four hub protein targets associated with SMF. Solute carrier family 16 member 1 (SLC16A1; also known as MCT1) encodes a transmembrane protein that enables proton-linked transport of a variety of monocarboxylate metabolites across the cellular membrane, including lactate, pyruvate, and ketone bodies (85–87). Lactate is a significant circulating carbohydrate fuel (88, 89). MCTs 1–4 (Slc16a1, Slc16a7, Slc16a3, and Slc16a4) regulate lactate entrance and departure from cells. These proteins' expression and activity may be modulated to affect lactate location in the body (90). Based on the existence of lactate shuttles, it is postulated that MCT1 expression may be significant for blood lactate elimination following supramaximal exercise, resulting in a better tolerance to muscle exhaustion (91, 92). GAA encodes the lysosomal enzyme acid alpha-glucosidase, which causes lysosomal glycogen buildup, swelling, and rupture in all tissues of the human body. Furthermore, autophagic accumulation, organelle abnormalities, and energy deficiencies are common (93). Pompe illness is characterized by a partial or total GAA deficiency, and in addition to symptoms associated with skeletal and respiratory muscle weakness, non-motor issues such as weariness can have a dramatic



Molecular docking of target genes with active ingredients of RSYRD. (A) Ferulic acid docking RB1. (B) Curcumin docking RB1. (C) Quercetin docking RB1. (D) Oleic acid docking RB1. (E) Acteoside docking RB1. (F) Curcumin docking NOS3. (G) Quercetin docking NOS3. (H) Isorhamnetin docking NOS3. (I) Benzoic acid docking NOS3. (J) Oleic acid docking NOS3. (K) Chlorogenic acid docking NOS3. (L) Ginsenoside Re docking NOS3. (M) Quercetin docking SLC16A1. (N) Benzoic acid docking SLC16A1. (O) Naringenin docking SLC16A1. (P) Pyruvic acid docking SLC16A1. (Q) Quercetin docking GAA. (R) Rutin docking GAA. (S) Isoquercitrin docking GAA. (T) (–)-Epicatechin docking GAA.

and devastating impact on the patient's life (94). Aval glucosidase alfa gained its initial approval in the United States in August 2021 for the treatment of patients 1 year of age and older with late-onset Pompe disease (GAA deficiency) (95, 96). The NOS3 gene has several genetic variations, including single nucleotide polymorphisms (SNPs), variable number of tandem repeats (VNTRs), microsatellites, and insertions/deletions (97). Endothelial nitric oxide synthase (NOS3) is an essential enzyme responsible for the synthesis of nitric oxide (NO) in the endothelial cells lining the blood vessels (98). Under normal physiological settings, NO is produced in cells by the enzyme nitric oxide synthase (NOS) converting l-arginine to l-citrulline (99). Nitric oxide regulates various physiological processes, such as muscle fiber



type, microtubule organization, fatigability, postexercise force recovery, and mitochondrial ATP synthesis efficiency, through cGMPdependent mechanisms (100, 101). The Retinoblastoma Susceptibility gene (RB1) was the initial human tumor suppressor gene to be discovered, and it plays a crucial role in the formation of retinoblastoma, a type of cancer that affects children's eyes (102). It could be employed as a biomarker for breast cancer cell sensitivity to GLUT1 inhibitors (103). These findings indicate that the RSYRD medication targets presented in this work are significantly linked to SMF and have a high medicinal potential, suggesting that Chinese medicine formulas targeting these genes for SMF could be developed.

The current study provides several notable advantages. First, this is the first study to employ MR to identify SMF therapeutic targets, based on data from the biggest publicly available SMF risk GWAS to date. Furthermore, network pharmacology, which is based on a "herbingredient-target" network, can expose small-molecule regulation principles in a high-throughput way and illuminate the complicated interaction between pharmacological targets and the material basis of RSYRD. The conclusive drug predictions showcase the medical potential of these genes, while the strong binding activity seen in molecular docking highlights their substantial value as drug targets. This study offers a thorough assessment, covering everything from identification to drug-binding qualities, and it provides solid evidence for four potential therapeutic targets for SMF. The method unveils the scientific foundation of TCM and creates a paradigm for understanding the molecular basis of Chinese medical formulas and predicting disease treatments. Additionally, additional research on RSYRD may have a significant impact on the development of novel targeted therapeutics for SMF, perhaps by the isolation and characterisation of the active chemicals in the efficacious Chinese medical formulae.

5 Limitations

Several limitations must be considered when evaluating this study. It is important to note that certain medications and target genes may not have been incorporated into the public database, mostly because of the inherent characteristics of network pharmacology research. The formulation of Chinese herbal medicine is intricate, and it remains uncertain whether there are synergistic benefits or negative effects among substances that are difficult to elucidate. Another limitation arises from the diverse composition of the study sample. Although the investigation of eQTLs includes individuals from non-European backgrounds, the SMF population consists exclusively of individuals of European descent. Due to differences in genetic background and patterns of linkage disequilibrium, variations in demographic backgrounds may introduce potential bias in calculations of the influence of Mendelian randomization. The study's generalizability is constrained by its overrepresentation of individuals of European descent. Further investigation and verification are required to extend the results to individuals of different ethnic backgrounds, in order to ensure the wider applicability of the findings. The accuracy of molecular docking analysis is greatly influenced by the quality of the protein structures and ligands. Although this strategy identifies potential drug targets, it does not guarantee their effectiveness in

real-world clinical scenarios. Additional experimental validation and clinical trials are necessary to determine the therapeutic potential of the identified targets.

6 Conclusion

In conclusion, MR analysis was used in this study to find potential drug targets for secondary malaise and fatigue. To discover the active components and anti-SMF targets of RSYRD, we employed the network pharmacology approach to generate a multi-dimensional network map. Our comprehensive study indicates a causal relationship between the risk of SMF and genetically inherited levels of circulating NOS3, GAA, IMPA1, P4HTM, RB1, and SLC16A1. The identified proteins may be appealing drug targets for SMF, especially NOS3, GAA, and SLC16A1. Furthermore, the medicinal efficacy of these targets was verified through the application of molecular docking and drug prediction. The results provide encouraging avenues for more potent SMF therapies, which could lower the cost of medication research and advance personalised medicine strategies.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary material, further inquiries can be directed to the corresponding authors.

Author contributions

FW: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft. LZ: Conceptualization, Data curation, Formal analysis, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft. HC: Conceptualization, Data curation, Formal analysis, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Resources, Software, Supervision, Validation, Visualization, Writing – original

References

1. Dukes JC, Chakan M, Mills A, Marcaurd M. Approach to fatigue: best practice. *Med Clin North Am.* (2021) 105:137–48. doi: 10.1016/j.mcna.2020.09.007

2. Latimer KM, Gunther A, Kopec M. Fatigue in adults: evaluation and management. *Am Fam Physician*. (2023) 108:58–69.

3. Strober LB. Fatigue in multiple sclerosis: a look at the role of poor sleep. *Front Neurol.* (2015) 6:21. doi: 10.3389/fneur.2015.00021

4. Skorvanek M, Nagyova I, Rosenberger J, Krokavcova M, Ghorbani Saeedian R, Groothoff JW, et al. Clinical determinants of primary and secondary fatigue in patients with Parkinson's disease. *J Neurol.* (2013) 260:1554–61. doi: 10.1007/s00415-012-6828-4

5. Van Herck M, Goertz YM, Lareau S. What is fatigue? Am J Respir Crit Care Med. (2023) 207:P1-2. doi: 10.1164/rccm.2075P1

6. Mustian KM, Alfano CM, Heckler C, Kleckner AS, Kleckner IR, Leach CR, et al. Comparison of pharmaceutical, psychological, and exercise treatments for cancerrelated fatigue: a meta-analysis. *JAMA Oncol.* (2017) 3:961–8. doi: 10.1001/ jamaoncol.2016.6914

7. Cramp F, Hewlett S, Almeida C, Kirwan JR, Choy EH, Chalder T, et al. Nonpharmacological interventions for fatigue in rheumatoid arthritis. *Cochrane Database Syst Rev.* (2013):CD008322. doi: 10.1002/14651858.CD008322.pub2 draft. SG: Writing – review & editing. JW: Writing – review & editing. AL: Conceptualization, Data curation, Formal analysis, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft. ZW: Conceptualization, Data curation, Formal analysis, Funding acquisition, Investigation, Methodology, Project administration, Resources, Software, Supervision, Validation, Visualization, Writing – original draft.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fnut.2024.1404123/ full#supplementary-material

8. Emerson C, Barhoun P, Olive L, Fuller-Tyszkiewicz M, Gibson PR, Skvarc D, et al. A systematic review of psychological treatments to manage fatigue in patients with inflammatory bowel disease. *J Psychosom Res.* (2021) 147:110524. doi: 10.1016/j. jpsychores.2021.110524

9. Liu J, Qiao B, Cai Y, Tan Z, Deng N. Diarrhea accompanies intestinal inflammation and intestinal mucosal microbiota dysbiosis during fatigue combined with a high-fat diet. *BMC Microbiol.* (2023) 23:151. doi: 10.1186/s12866-023-02896-9

 Ruiter AM, Verschuuren J, Tannemaat MR. Fatigue in patients with myasthenia gravis. A systematic review of the literature. *Neuromuscul Disord*. (2020) 30:631–9. doi: 10.1016/j.nmd.2020.06.010

11. Chen Y, Xu S, Shen J, Yang H, Xu W, Shao M, et al. Effect of exercise on fatigue in multiple sclerosis patients: a network meta-analysis. *Int J Sports Med.* (2021) 42:1250–9. doi: 10.1055/a-1524-1935

12. Estevez-Lopez F, Maestre-Cascales C, Russell D, Alvarez-Gallardo IC, Rodriguez-Ayllon M, Hughes CM, et al. Effectiveness of exercise on fatigue and sleep quality in fibromyalgia: a systematic review and meta-analysis of randomized trials. *Arch Phys Med Rehabil.* (2021) 102:752–61. doi: 10.1016/j.apmr.2020.06.019

13. Merriwether EN, Frey-Law LA, Rakel BA, Zimmerman MB, Dailey DL, Vance CGT, et al. Physical activity is related to function and fatigue but not pain in women

with fibromyalgia: baseline analyses from the fibromyalgia activity study with TENS (FAST). *Arthritis Res Ther.* (2018) 20:199. doi: 10.1186/s13075-018-1671-3

14. Wang C, Schmid CH, Fielding RA, Harvey WF, Reid KF, Price LL, et al. Effect of tai chi versus aerobic exercise for fibromyalgia: comparative effectiveness randomized controlled trial. *BMJ.* (2018) 360:k851. doi: 10.1136/bmj.k851

15. Zhang Y, Lin L, Li H, Hu Y, Tian L. Effects of acupuncture on cancer-related fatigue: a meta-analysis. *Support Care Cancer*. (2018) 26:415–25. doi: 10.1007/s00520-017-3955-6

16. Kinkead B, Schettler PJ, Larson ER, Carroll D, Sharenko M, Nettles J, et al. Massage therapy decreases cancer-related fatigue: results from a randomized early phase trial. *Cancer*. (2018) 124:546–54. doi: 10.1002/cncr.31064

17. Qiao B, Liu J, Peng X, Cai Y, Peng M, Li X, et al. Association of short-chain fatty acids with gut microbiota and lipid metabolism in mice with diarrhea induced by high-fat diet in a fatigued state. *Mol Nutr Food Res.* (2023) 67:e2300452. doi: 10.1002/mnfr.202300452

18. Arends J, Strasser F, Gonella S, Solheim TS, Madeddu C, Ravasco P, et al. Cancer cachexia in adult patients: ESMO clinical practice guidelines. *ESMO Open.* (2021) 6:100092. doi: 10.1016/j.esmoop.2021.100092

19. Nourbakhsh B, Revirajan N, Morris B, Cordano C, Creasman J, Manguinao M, et al. Safety and efficacy of amantadine, modafinil, and methylphenidate for fatigue in multiple sclerosis: a randomised, placebo-controlled, crossover, double-blind trial. *Lancet Neurol.* (2021) 20:38–48. doi: 10.1016/S1474-4422(20)30354-9

20. Yagi T, Sawada K, Miyamoto M, Kinose Y, Nakagawa S, Takiuchi T, et al. Safety and efficacy of Ninjin'yoeito along with iron supplementation therapy for preoperative anemia, fatigue, and anxiety in patients with gynecological disease: an open-label, single-center, randomized phase-II trial. *BMC Womens Health*. (2022) 22:229. doi: 10.1186/s12905-022-01824-9

21. Sato S. Apathy and fatigue in autistic spectrum disorder improved by Japanese herbal medicine: TSUMURA Ninjin'yoeito extract. *Psychiatry Clin Neurosci.* (2018) 72:540. doi: 10.1111/pcn.12667

22. Fukuhara K, Edagawa E, Nishioka T, Shinjo S. The effect of Ninjinyoeito for patients undergoing chemotherapy. *Gan To Kagaku Ryoho*. (2019) 46:1033–7.

23. Ito T, Konishi A, Tsubokura Y, Azuma Y, Hotta M, Yoshimura H, et al. Combined use of Ninjin'yoeito improves subjective fatigue caused by lenalidomide in patients with multiple myeloma: a retrospective study. *Front Nutr.* (2018) 5:72. doi: 10.3389/ fnut.2018.00072

24. Okada KI, Kawai M, Hirono S, Miyazawa M, Kitahata Y, Kobayashi R, et al. Impact of Ninjin'yoeito on fatigue in patients receiving nab-paclitaxel plus gemcitabine therapy: a prospective, single-arm, phase II open label, nonrandomized, historically-controlled study. *Curr Ther Res Clin Exp.* (2020) 93:100605. doi: 10.1016/j.curtheres.2020.100605

25. Xu Y, Chen Y, Li P, Wang XS. Ren Shen Yangrong Tang for fatigue in cancer survivors: a phase I/II open-label study. *J Altern Complement Med.* (2015) 21:281–7. doi: 10.1089/acm.2014.0211

26. Suzuki S, Aihara F, Shibahara M, Sakai K. Safety and effectiveness of Ninjin'yoeito: a utilization study in elderly patients. *Front Nutr.* (2019) 6:14. doi: 10.3389/fnut.2019.00014

27. Xu Y, Wang XS, Chen Y, Shi Q, Chen TH, Li P. A phase II randomized controlled trial of Renshen Yangrong Tang herbal extract granules for fatigue reduction in cancer survivors. *J Pain Symptom Manag.* (2020) 59:966–73. doi: 10.1016/j.jpainsymman.2019.10.018

28. Kawabe M, Hayashi A, Komatsu M, Inui A, Shiozaki K. Ninjinyoeito improves anxiety behavior in neuropeptide Y deficient zebrafish. *Neuropeptides*. (2021) 87:102136. doi: 10.1016/j.npep.2021.102136

29. Murata K, Fujita N, Takahashi R, Inui A. Ninjinyoeito improves behavioral abnormalities and hippocampal neurogenesis in the corticosterone model of depression. *Front Pharmacol.* (2018) 9:1216. doi: 10.3389/fphar.2018.01216

30. Oizumi H, Imai R, Suzuki T, Omiya Y, Tanaka KF, Mimura M, et al. Ninjin'yoeito, a traditional Japanese Kampo medicine, suppresses the onset of anhedonia induced by dysfunction in the striatal dopamine receptor type 2-expressing medium spiny neurons. *Neuroreport.* (2021) 32:869–74. doi: 10.1097/WNR.00000000001667

31. Kudoh C, Arita R, Honda M, Kishi T, Komatsu Y, Asou H, et al. Effect of Ninjin'yoeito, a Kampo (traditional Japanese) medicine, on cognitive impairment and depression in patients with Alzheimer's disease: 2 years of observation. *Psychogeriatrics*. (2016) 16:85–92. doi: 10.1111/psyg.12125

32. King EA, Davis JW, Degner JF. Are drug targets with genetic support twice as likely to be approved? Revised estimates of the impact of genetic support for drug mechanisms on the probability of drug approval. *PLoS Genet*. (2019) 15:e1008489. doi: 10.1371/ journal.pgen.1008489

33. Hingorani AD, Kuan V, Finan C, Kruger FA, Gaulton A, Chopade S, et al. Improving the odds of drug development success through human genomics: modelling study. *Sci Rep.* (2019) 9:18911. doi: 10.1038/s41598-019-54849-w

34. Schmidt AF, Finan C, Gordillo-Maranon M, Asselbergs FW, Freitag DF, Patel RS, et al. Genetic drug target validation using Mendelian randomisation. *Nat Commun.* (2020) 11:3255. doi: 10.1038/s41467-020-16969-0

35. Finan C, Gaulton A, Kruger FA, Lumbers RT, Shah T, Engmann J, et al. The druggable genome and support for target identification and validation in drug development. *Sci Transl Med.* (2017) 9:eaag1166. doi: 10.1126/scitranslmed.aag1166

36. Ishigaki K, Sakaue S, Terao C, Luo Y, Sonehara K, Yamaguchi K, et al. Multiancestry genome-wide association analyses identify novel genetic mechanisms in rheumatoid arthritis. *Nat Genet*. (2022) 54:1640–51. doi: 10.1038/s41588-022-01213-w

37. Chauquet S, Zhu Z, O'Donovan MC, Walters JTR, Wray NR, Shah S. Association of antihypertensive drug target genes with psychiatric disorders: a Mendelian randomization study. *JAMA Psychiatry*. (2021) 78:623–31. doi: 10.1001/jamapsychiatry.2021.0005

38. Wang Q, Oliver-Williams C, Raitakari OT, Viikari J, Lehtimaki T, Kahonen M, et al. Metabolic profiling of angiopoietin-like protein 3 and 4 inhibition: a drug-target Mendelian randomization analysis. *Eur Heart J.* (2021) 42:1160–9. doi: 10.1093/ eurheartj/ehaa972

39. Nogales C, Mamdouh ZM, List M, Kiel C, Casas AI, Schmidt H. Network pharmacology: curing causal mechanisms instead of treating symptoms. *Trends Pharmacol Sci.* (2022) 43:136–50. doi: 10.1016/j.tips.2021.11.004

40. Pinzi I., Rastelli G. Molecular docking: shifting paradigms in drug discovery. Int J Mol Sci. (2019) 20:4331. doi: 10.3390/ijms20184331

41. Kong X, Liu C, Zhang Z, Cheng M, Mei Z, Li X, et al. BATMAN-TCM 2.0: an enhanced integrative database for known and predicted interactions between traditional Chinese medicine ingredients and target proteins. *Nucleic Acids Res.* (2023) 52:D1110–20. doi: 10.1093/nar/gkad926

42. Ru J, Li P, Wang J, Zhou W, Li B, Huang C, et al. TCMSP: a database of systems pharmacology for drug discovery from herbal medicines. *J Cheminform*. (2014) 6:13. doi: 10.1186/1758-2946-6-13

43. Vosa U, Claringbould A, Westra HJ, Bonder MJ, Deelen P, Zeng B, et al. Large-scale cis- and trans-eQTL analyses identify thousands of genetic loci and polygenic scores that regulate blood gene expression. *Nat Genet.* (2021) 53:1300–10. doi: 10.1038/ s41588-021-00913-z

44. Su WM, Gu XJ, Dou M, Duan QQ, Jiang Z, Yin KF, et al. Systematic druggable genome-wide Mendelian randomisation identifies therapeutic targets for Alzheimer's disease. *J Neurol Neurosurg Psychiatry*. (2023) 94:954–61. doi: 10.1136/jnnp-2023-331142

45. Verbanck M, Chen CY, Neale B, Do R. Detection of widespread horizontal pleiotropy in causal relationships inferred from Mendelian randomization between complex traits and diseases. *Nat Genet*. (2018) 50:693–8. doi: 10.1038/s41588-018-0099-7

46. Wang XF, Xu WJ, Wang FF, Leng R, Yang XK, Ling HZ, et al. Telomere length and development of systemic lupus erythematosus: a Mendelian randomization study. *Arthritis Rheumatol.* (2022) 74:1984–90. doi: 10.1002/art.42304

47. Hemani G, Zheng J, Elsworth B, Wade KH, Haberland V, Baird D, et al. The MRbase platform supports systematic causal inference across the human phenome. *eLife*. (2018) 7:e34408. doi: 10.7554/eLife.34408

48. Otasek D, Morris JH, Boucas J, Pico AR, Demchak B. Cytoscape automation: empowering workflow-based network analysis. *Genome Biol.* (2019) 20:185. doi: 10.1186/s13059-019-1758-4

49. Yu G, Wang LG, Han Y, He QY. clusterProfiler: an R package for comparing biological themes among gene clusters. *OMICS*. (2012) 16:284–7. doi: 10.1089/ omi.2011.0118

50. Luo W, Brouwer C. Pathview: an R/Bioconductor package for pathway-based data integration and visualization. *Bioinformatics*. (2013) 29:1830–1. doi: 10.1093/bioinformatics/btt285

51. Yoo M, Shin J, Kim J, Ryall KA, Lee K, Lee S, et al. DSigDB: drug signatures database for gene set analysis. *Bioinformatics*. (2015) 31:3069–71. doi: 10.1093/bioinformatics/btv313

52. Kuleshov MV, Jones MR, Rouillard AD, Fernandez NF, Duan Q, Wang Z, et al. Enrichr: a comprehensive gene set enrichment analysis web server 2016 update. *Nucleic Acids Res.* (2016) 44:W90–7. doi: 10.1093/nar/gkw377

53. Ru W, Wang D, Xu Y, He X, Sun YE, Qian L, et al. Chemical constituents and bioactivities of *Panax ginseng* (C. A. Mey.). *Drug Discov Ther.* (2015) 9:23–32. doi: 10.5582/ddt.2015.01004

54. Liu C, Wang S, Xiang Z, Xu T, He M, Xue Q, et al. The chemistry and efficacy benefits of polysaccharides from *Atractylodes macrocephala* Koidz. *Front Pharmacol.* (2022) 13:952061. doi: 10.3389/fphar.2022.952061

55. Ahn KI, Choi EO, Kwon DH, Hwang Bo H, Kim MY, Kim HJ, et al. Induction of apoptosis by ethanol extract of *Citrus unshiu* Markovich peel in human bladder cancer T24 cells through ROS-mediated inactivation of the PI3K/Akt pathway. *Biosci Trends*. (2017) 11:565–73. doi: 10.5582/bst.2017.01218

56. Zou J, Wang J, Ye W, Lu J, Li C, Zhang D, et al. Citri Reticulatae Pericarpium (Chenpi): a multi-efficacy pericarp in treating cardiovascular diseases. *Biomed Pharmacother*. (2022) 154:113626. doi: 10.1016/j.biopha.2022.113626

57. Ding Y, Brand E, Wang W, Zhao Z. Licorice: resources, applications in ancient and modern times. J Ethnopharmacol. (2022) 298:115594. doi: 10.1016/j.jep.2022.115594

58. Nie A, Chao Y, Zhang X, Jia W, Zhou Z, Zhu C. Phytochemistry and pharmacological activities of Wolfiporia Cocos (F.A. Wolf) Ryvarden & Gilb. Front Pharmacol. (2020) 11:505249. doi: 10.3389/fphar.2020.505249

59. Liu P, Zhao H, Luo Y. Anti-aging implications of Astragalus membranaceus (Huangqi): a well-known Chinese tonic. Aging Dis. (2017) 8:868–86. doi: 10.14336/ AD.2017.0816 60. Bae JY, Kim CY, Kim HJ, Park JH, Ahn MJ. Differences in the chemical profiles and biological activities of *Paeonia lactiflora* and *Paeonia obovata*. J Med Food. (2015) 18:224–32. doi: 10.1089/jmf.2014.3144

61. Zhou YM, Dong XR, Xu D, Tang J, Cui YL. Therapeutic potential of traditional Chinese medicine for interstitial lung disease. *J Ethnopharmacol.* (2024) 318:116952. doi: 10.1016/j.jep.2023.116952

62. Jia J, Chen J, Wang G, Li M, Zheng Q, Li D. Progress of research into the pharmacological effect and clinical application of the traditional Chinese medicine Rehmanniae radix. *Biomed Pharmacother*. (2023) 168:115809. doi: 10.1016/j. biopha.2023.115809

63. Kopustinskiene DM, Bernatoniene J. Antioxidant effects of *Schisandra chinensis* fruits and their active constituents. *Antioxidants.* (2021) 10:620. doi: 10.3390/antiox10040620

64. Bhattarai N, Kumbhar AA, Pokharel YR, Yadav PN. Anticancer potential of coumarin and its derivatives. *Mini Rev Med Chem.* (2021) 21:2996–3029. doi: 10.217 4/1389557521666210405160323

65. Zeng W, Wu AG, Zhou XG, Khan I, Zhang RL, Lo HH, et al. Saponins isolated from radix polygalae extent lifespan by modulating complement C3 and gut microbiota. *Pharmacol Res.* (2021) 170:105697. doi: 10.1016/j.phrs.2021.105697

66. Xiang H, Zuo J, Guo F, Dong D. What we already know about rhubarb: a comprehensive review. *Chin Med.* (2020) 15:88. doi: 10.1186/s13020-020-00370-6

67. Agaimy A. Paraneoplastic disorders associated with miscellaneous neoplasms with focus on selected soft tissue and undifferentiated/rhabdoid malignancies. *Semin Diagn Pathol.* (2019) 36:269–78. doi: 10.1053/j.semdp.2019.02.001

68. Qiao B, Liu J, Deng N, Cai Y, Bian Y, Wu Y, et al. Gut content microbiota dysbiosis and dysregulated lipid metabolism in diarrhea caused by high-fat diet in a fatigued state. *Food Funct*. (2023) 14:3880–92. doi: 10.1039/D3FO00378G

69. Dong J, Wang S, Gui Y, Wang D, Ma X, Hu S, et al. Astragalus membranaceus (Huang Qi) for cancer-related fatigue: a protocol for systematic review and metaanalysis. Medicine. (2022) 101:e28633. doi: 10.1097/MD.00000000028633

70. Liu CH, Tsai CH, Li TC, Yang YW, Huang WS, Lu MK, et al. Effects of the traditional Chinese herb *Astragalus membranaceus* in patients with poststroke fatigue: a double-blind, randomized, controlled preliminary study. *J Ethnopharmacol.* (2016) 194:954–62. doi: 10.1016/j.jep.2016.10.058

71. Xu L, Xu XY, Hou XQ, Wang FG, Gao S, Zhang HT. Adjuvant therapy with *Astragalus membranaceus* for post-stroke fatigue: a systematic review. *Metab Brain Dis.* (2020) 35:83–93. doi: 10.1007/s11011-019-00483-4

72. Arring NM, Millstine D, Marks LA, Nail LM. Ginseng as a treatment for fatigue: a systematic review. *J Altern Complement Med.* (2018) 24:624–33. doi: 10.1089/acm.2017.0361

73. Bach HV, Kim J, Myung SK, Cho YA. Efficacy of ginseng supplements on fatigue and physical performance: a meta-analysis. *J Korean Med Sci.* (2016) 31:1879–86. doi: 10.3346/jkms.2016.31.12.1879

74. Yennurajalingam S, Tannir NM, Williams JL, Lu Z, Hess KR, Frisbee-Hume S, et al. Randomized, placebo-controlled trial of *Panax ginseng* for cancer-related fatigue in patients with advanced cancer. *J Natl Compr Cancer Netw*. (2017) 15:1111–20. doi: 10.6004/jnccn.2017.0149

75. Guglielmo M, Di Pede P, Alfieri S, Bergamini C, Platini F, Ripamonti CI, et al. A randomized, double-blind, placebo controlled, phase II study to evaluate the efficacy of ginseng in reducing fatigue in patients treated for head and neck cancer. *J Cancer Res Clin Oncol.* (2020) 146:2479–87. doi: 10.1007/s00432-020-03300-z

76. Li X, Yang M, Zhang YL, Hou YN, Smith CM, Korenstein D, et al. Ginseng and ginseng herbal formulas for symptomatic Management of Fatigue: a systematic review and meta-analysis. *J Integr Complement Med.* (2023) 29:468–82. doi: 10.1089/jicm.2022.0532

77. Luo C, Xu X, Wei X, Feng W, Huang H, Liu H, et al. Natural medicines for the treatment of fatigue: bioactive components, pharmacology, and mechanisms. *Pharmacol Res.* (2019) 148:104409. doi: 10.1016/j.phrs.2019.104409

78. Rondanelli M, Riva A, Petrangolini G, Gasparri C, Perna S. Two-month period of 500 mg lecithin-based delivery form of quercetin daily dietary supplementation counterbalances chronic fatigue symptoms: a double-blind placebo controlled clinical trial. *Biomed Pharmacother*. (2023) 167:115453. doi: 10.1016/j.biopha.2023.115453

79. Saghatelyan T, Tananyan A, Janoyan N, Tadevosyan A, Petrosyan H, Hovhannisyan A, et al. Efficacy and safety of curcumin in combination with paclitaxel in patients with advanced, metastatic breast cancer: a comparative, randomized, double-blind, placebocontrolled clinical trial. *Phytomedicine*. (2020) 70:153218. doi: 10.1016/j. phymed.2020.153218

80. Chung TH, Kim JH, Seol SY, Kim YJ, Lee YJ. The effects of Korean red ginseng on biological aging and antioxidant capacity in postmenopausal women: a double-blind randomized controlled study. *Nutrients*. (2021) 13:3090. doi: 10.3390/nu13093090

81. Su KY, Yu CY, Chen YW, Huang YT, Chen CT, Wu HF, et al. Rutin, a flavonoid and principal component of *Saussurea involucrata*, attenuates physical fatigue in a forced swimming mouse model. *Int J Med Sci.* (2014) 11:528–37. doi: 10.7150/ijms.8220

82. Ren JL, Yang L, Qiu S, Zhang AH, Wang XJ. Efficacy evaluation, active ingredients, and multitarget exploration of herbal medicine. *Trends Endocrinol Metab.* (2023) 34:146–57. doi: 10.1016/j.tem.2023.01.005

83. Poornima P, Kumar JD, Zhao Q, Blunder M, Efferth T. Network pharmacology of cancer: from understanding of complex interactomes to the design of multi-target specific therapeutics from nature. *Pharmacol Res.* (2016) 111:290–302. doi: 10.1016/j. phrs.2016.06.018

84. Zhang BY, Zheng YF, Zhao J, Kang D, Wang Z, Xu LJ, et al. Identification of multitarget anti-cancer agents from TCM formula by *in silico* prediction and in vitro validation. *Chin J Nat Med.* (2022) 20:332–51. doi: 10.1016/S1875-5364(22)60180-8

85. van Hasselt PM, Ferdinandusse S, Monroe GR, Ruiter JP, Turkenburg M, Geerlings MJ, et al. Monocarboxylate transporter 1 deficiency and ketone utilization. *N Engl J Med.* (2014) 371:1900–7. doi: 10.1056/NEJMoa1407778

86. Jha MK, Passero JV, Rawat A, Ament XH, Yang F, Vidensky S, et al. Macrophage monocarboxylate transporter 1 promotes peripheral nerve regeneration after injury in mice. *J Clin Invest*. (2021) 131:e141964. doi: 10.1172/JCI141964

87. Eichner R, Heider M, Fernandez-Saiz V, van Bebber F, Garz AK, Lemeer S, et al. Immunomodulatory drugs disrupt the cereblon-CD147-MCT1 axis to exert antitumor activity and teratogenicity. *Nat Med.* (2016) 22:735–43. doi: 10.1038/nm.4128

88. Brooks GA. The science and translation of lactate shuttle theory. *Cell Metab.* (2018) 27:757–85. doi: 10.1016/j.cmet.2018.03.008

89. Li X, Yang Y, Zhang B, Lin X, Fu X, An Y, et al. Lactate metabolism in human health and disease. *Signal Transduct Target Ther.* (2022) 7:305. doi: 10.1038/ s41392-022-01151-3

90. Rabinowitz JD, Enerback S. Lactate: the ugly duckling of energy metabolism. *Nat Metab.* (2020) 2:566–71. doi: 10.1038/s42255-020-0243-4

91. Thomas C, Perrey S, Lambert K, Hugon G, Mornet D, Mercier J. Monocarboxylate transporters, blood lactate removal after supramaximal exercise, and fatigue indexes in humans. *J Appl Physiol.* (1985) 98:804–9. doi: 10.1152/japplphysiol.01057.2004

92. Liu J, Qiao B, Deng N, Wu Y, Li D, Tan Z. The diarrheal mechanism of mice with a high-fat diet in a fatigued state is associated with intestinal mucosa microbiota. *3 Biotech.* (2023) 13:77. doi: 10.1007/s13205-023-03491-5

93. Schoser B. Pompe disease: what are we missing? Ann Transl Med. (2019) 7:292. doi: 10.21037/atm.2019.05.29

94. Hagemans ML, van Schie SP, Janssens AC, van Doorn PA, Reuser AJ, van der Ploeg AT. Fatigue: an important feature of late-onset Pompe disease. *J Neurol.* (2007) 254:941–5. doi: 10.1007/s00415-006-0434-2

95. Dhillon S. Avalglucosidase alfa: first Approval. Drugs. (2021) 81:1803-9. doi: 10.1007/s40265-021-01600-3

96. Scheffers LE, Somers OC, Dulfer K, Dieleman GC, Walet S, van der Giessen LJ, et al. Physical training and high-protein diet improved muscle strength, parent-reported fatigue, and physical quality of life in children with Pompe disease. *J Inherit Metab Dis.* (2023) 46:605–17. doi: 10.1002/jimd.12607

97. Oliveira-Paula GH, Lacchini R, Tanus-Santos JE. Clinical and pharmacogenetic impact of endothelial nitric oxide synthase polymorphisms on cardiovascular diseases. *Nitric Oxide*. (2017) 63:39–51. doi: 10.1016/j.niox.2016.08.004

98. Luizon MR, Pereira DA, Tanus-Santos JE. Pharmacogenetic relevance of endothelial nitric oxide synthase polymorphisms and gene interactions. *Pharmacogenomics.* (2018) 19:1423–35. doi: 10.2217/pgs-2018-0098

99. Vannini F, Kashfi K, Nath N. The dual role of iNOS in cancer. *Redox Biol.* (2015) 6:334–43. doi: 10.1016/j.redox.2015.08.009

100. Moon Y, Balke JE, Madorma D, Siegel MP, Knowels G, Brouckaert P, et al. Nitric oxide regulates skeletal muscle fatigue, fiber type, microtubule organization, and mitochondrial ATP synthesis efficiency through cGMP-dependent mechanisms. *Antioxid Redox Signal.* (2017) 26:966–85. doi: 10.1089/ars.2016.6630

101. Heydemann A, McNally E. NO more muscle fatigue. J Clin Invest. (2009) 119:448-50. doi: 10.1172/JCI38618

102. Yao Y, Gu X, Xu X, Ge S, Jia R. Novel insights into RB1 mutation. *Cancer Lett.* (2022) 547:215870. doi: 10.1016/j.canlet.2022.215870

103. Wu Q, Ba-Alawi W, Deblois G, Cruickshank J, Duan S, Lima-Fernandes E, et al. GLUT1 inhibition blocks growth of RB1-positive triple negative breast cancer. *Nat Commun.* (2020) 11:4205. doi: 10.1038/s41467-020-18020-8