

**Research Paper** 



2019; 10(26): 6526-6534. doi: 10.7150/jca.33666

# Bioinformatics analysis of differentially expressed genes in hepatocellular carcinoma cells exposed to Swertiamarin

Haoran Tang<sup>1\*</sup>, Yang Ke<sup>2\*</sup>, Zongfang Ren<sup>3\*</sup>, Xuefen Lei<sup>4\*</sup>, Shufeng Xiao<sup>2\*</sup>, Tianhao Bao<sup>5\*</sup>, Zhitian Shi<sup>2</sup>, Renchao Zou<sup>2</sup>, Tiangen Wu<sup>2</sup>, Jian Zhou<sup>2</sup>, Chang-An Geng<sup>6</sup>, Lin Wang<sup>2⊠</sup>, Jijun Chen<sup>6⊠</sup>

- 1. Department of Gastroenterological Surgery, the Second Affiliated Hospital of Kunming Medical University, Kunming, China.
- 2. Department of Hepatobiliary Surgery, the Second Affiliated Hospital of Kunming Medical University, Kunming, China.
- 3. Department of Critical Care Medicine, the Second Affiliated Hospital of Kunming Medical University, Kunming, China.
- 4. Department of Oncology, the Second Affiliated Hospital of Kunming Medical University, Kunming, China.
- 5. Mental Health Center, Kunming Medical University, Kunming, China.
- 6. State Key Laboratory of Phytochemistry and Plant Resources in West China, Kunming Institute of Botany, Chinese Academy of Sciences, Kunming 650201, PR China.

\* These authors share the first authorship.

Corresponding authors: Dr. Lin Wang, Kunming Medical University, No. 1168 Chunrongxi Road, Chenggong District, Kunming 650500, Yunnan, China. Tel: +86-871-65352087. E-mail: linwang0705@126.com; Dr. Jijun Chen, the State Key Laboratory of Phytochemistry and Plant Resources in West China, Kunming Institute of Botany, Chinese Academy of Sciences, Kunming 650201, PR China; Tel: +86-871-5223265. E-mail: chenjj@mail.kib.ac.cn

© The author(s). This is an open access article distributed under the terms of the Creative Commons Attribution License (https://creativecommons.org/licenses/by/4.0/). See http://ivyspring.com/terms for full terms and conditions.

Received: 2019.01.30; Accepted: 2019.09.22; Published: 2019.10.21

#### Abstract

Aim: To explore gene expression profiling in hepatocellular carcinoma (HCC) cells exposed to swertiamarin.

**Methods**: Cell viability, apoptosis and invasion were examined in HepG2 cells after swertiamarin treatment. Tumor growth of SK-Hep-1 cells xenografted in nude mice was monitored after swertiamarin treatment. Total RNA was isolated from HepG2 cells treated with swertiamarin for microarray analysis. The data of microarray were analyzed by bioinformatics.

**Results:** Swertiamarin treatment decreased the viability and invasion while increased the apoptosis of HepG2 cells, and significantly inhibited the growth of SK-Hep-1 cells xenografted in nude mice. Pathway and biological process analysis of differentially expressed genes (DEGs) in swertiamarin treated HepG2 cells showed that Pl3k-Akt was the most significant regulated pathway. 47 targets of swertiamarin were predicted by CGBVS while 21 targets were predicted by 3NN. Notably, 8 targets were predicted as the targets of swertiamarin by both programs, including two prominent targets JUN and STAT3. A large range of DEGs induced by swertiamarin could be regulated by JUN and STAT3.

**Conclusion**: Swertiamarin treatment led to significant changes in the expression of a variety of genes that modulate cell survival, cell cycle progression, apoptosis, and invasion. Moreover, most of these genes can be clustered into pathway networks such as PI3K, JUN, STAT3, which are predicted targets of swertiamarin. Further confirmation of these targets will reveal the anti-tumor mechanisms of swertiamarin and facilitate the development of swertiamarin as a novel agent for cancer prevention and treatment.

Key words: swertiamarin; microarray; HepG2 cells; hepatocellular cancer; Jun; Stat3

## Introduction

Hepatocellular carcinoma (HCC) is a common cancer in China and becomes the fourth leading cause of cancer related deaths worldwide [1]. Despite recent developments in surgery, radiotherapy, and chemotherapy, current therapeutic approaches for HCC are still not effective. Recently, great efforts have been taken to investigate herbs with potential anti-cancer efficacy [2-4]. Nevertheless, herb extracts are usually crude mixture of many compounds, and the complexity of herbs hinders the development of anti-cancer drugs. Therefore, it is important to isolate, identify and elucidate the mechanism of action of single compound from herbs as lead candidate for cancer drug discovery.

Swertiamarin (STM) is a main bioactive component in Swertia mussotii Franch (Gentianaceae), a commonly used Tibetan medicine, exhibits broad bioactivities and including hepatoprotective effects [5-7]. In rat model of liver d-Galactosamine, injury induced bv oral administration of STM caused significant antioxidant and hepatoprotective effects [5]. Total iridoids and xanthones (TIXS) were extracted from Swertia mussotii Franch, and STM was identified as the main component of TIXS. In mouse model of alpha-naphthylisot hiocyanate induced liver damage, oral administration of TIXS significantly reduced levels of alanine aminotransferase, aspartate aminotransferase and the total and direct bilirubins [6]. Furthermore, in rat model of liver injury induced by bile duct ligation, treatment with STM led to significantly decreased serum levels of serum alanine aminotransferase and aspartate aminotransferase, significantly improved liver histology, and significantly reduced inflammation and cholestasis [7]. However, the efficacy of STM on HCC treatment has not been investigated.

The evaluation of anti-HCC efficacy of STM will greatly help develop novel therapeutics for HCC. To better understand the mechanism of action of STM in HCC therapy, in this study we employed microarray to screen differentially expressed genes in HepG2 hepatocellular carcinoma cells treated by STM.

## Materials and Methods

## Cell culture

HepG<sub>2</sub> and SK-Hep-1 cells were cultured in Dulbecco's modified Eagle's Medium (DMEM) supplemented with 10% fetal bovine serum and 1% penicillin and streptomycin at 37°C in a humid incubator with 5% CO<sub>2</sub>. STM (>95% purity) was purchased from Sigma Aldrich (St. Louis, MO, USA) and dissolved in dimethylsulfoxide (DMSO) to make stock solution at 60 mg/ml.

## Nude mouse model

Male BALB/c-nu nude mice (weight 18-22 g) were purchased from Animal Center of Kunming Medical University (No. SCXK-2015-0002) and animal experiments were approved by Animal Care and Use Committee of Kunming Medical University. Each

mouse received subcutaneous injection of  $3 \times 10^6$  SK-Hep-1 cells, and when the tumor grew the mice were randomly divided into 2 groups (n=6): control group received intratumoral injection of PBS; STM group received intratumoral injection of 5 µg STM. Tumor volume was measured every week for 8 weeks, and then the mice were sacrificed and xenograft tumors were dissected to measure the weight.

## Cell viability assay

HepG<sub>2</sub> cells were seeded into 96-well plates at a density of 5,000 cells/well. After overnight incubation in a humid chamber at 37°C, the cells were treated with DMSO or different dose of STM. Viable cells were evaluated using CCK-8 Assay kit (Dojindo, Japan) according to the manufacturer's instructions. CCK-8 solution was added to the cells in 96-well plates and the plates were incubated at 37°C for 4 h, and then 150  $\mu$ L DMSO was added to each well and the plates were incubated at room temperature for 10 min. The optical density of each well was read at 450 nm using a microplate reader (Bio-Rad, Hercules, CA, USA).

## TUNEL assay

HepG<sub>2</sub> cells were seeded onto coverslips in 6-well plates. After overnight incubation in a humid chamber at 37°C, the cells were treated with DMSO or STM. The cells were fixed in 4% paraformaldehyde for 15 min, and then apoptotic cells were stained using TUNEL kit (Roche Diagnostic, Indianapolis, IN, USA) following the manufacturer's instructions. Apoptotic cells showed brown nuclear staining.

## In vitro cell invasion assay

Transwell chamber (Costar, Cambridge, MA, USA) was used to evaluate cell invasion. The upper chambers were precoated with Matrigel overnight at  $4^{\circ}$ C. HepG<sub>2</sub> cells were treated with DMSO or STM, and 100 uL of the cell suspensions were seeded in the upper chambers, while the lower chambers were filled with complete medium supplemented with 10% BSA. After 24 h incubation, the cells on the upper surface of the filters were wiped by cotton swabs and the cells on the underside of the filters were fixed, stained with crystal violet and observed under a microscope.

## Microarray

After HepG2 cells were treated with DMSO or STM, total RNA from each sample was isolated using Trizol (Invitrogen) and purified using mirVana miRNA Isolation Kit (Ambion, Austin, TX, USA) following the manufacturer's instructions. RNA integrity was determined by denatured agarose gel electrophoresis. DNA microarray was performed according to standard protocol.



**Figure 1. STM inhibited HCC.** A. Chemical structure and the origin of STM. B. STM decreased the viability of HepG2 cells in a dose-dependent manner. C. STM increased the apoptosis of HepG2 cells. D. STM decreased the invasion of HepG2 cells. D. STM decreased the invasion of HepG2 cells. D. STM decreased the invasion of HepG2 cells. The data were presented as mean  $\pm$  SD (n=3). \*\*\* P<0.001. Scale bar: 20 µm. E. Tumor tissues dissected from nude mice which received subcutaneous injection of SK-Hep-1 cells. F. Tumor growth curves of each group of nude mice which received subcutaneous injection of SK-Hep-1 cells. G. The weight of tumor tissues dissected from nude mice which received subcutaneous injection of SK-Hep-1 cells. The data were presented as mean  $\pm$  SD (n=6). \*P<0.05, \*\*P<0.01, \*\*\*P<0.001.To select the differentially expressed genes, we used threshold values of  $\geq$ 2 and  $\leq$ -2-fold change and Benjamini-Hochberg corrected p value of 0.05. The data were Log2 transformed and median centered by genes using the Adjust Data function of CLUSTER 3.0 software. Finally, tree visualization was performed using Java Treeview (Stanford University School of Medicine, Stanford, CA, USA).

CapitalBio cRNA Amplification and Labeling Kit (CapitalBio) was used to produce fluorescent dye labeled cDNA. The labeled cDNAs were then hybridized to Agilent human mRNA Array which was designed with eight identical arrays per slide (8 x 60K format). After hybridization at 42°C overnight, the arrays were washed with 2X SSC and 0.2% SDS at 42°C for 5 min, followed by washing with 0.2X SSC and 0.2% SDS at room temperature for 5 min. The array data were analyzed for data summarization, normalization and quality control by using the GeneSpring software V13 (Agilent).

#### Statistical analysis

The data were presented as mean ± standard deviation (SD). Data from two samples were compared using t test and data from multiple samples were compared using single factor analysis of variance. P<0.05 was considered statistically significant.

#### Results

#### **STM** inhibits HCC

The origin and the structure of STM were shown in Fig. 1A. STM inhibited the viability of HepG2 cells in a dose dependent manner (Fig. 1B). Moreover, STM significantly increased the apoptosis of HepG2 cells compared to vehicle treated cells (Fig. 1C). Furthermore, STM significantly inhibited the invasion of HepG2 cells compared to vehicle treated cells (Fig. 1D). To further confirm that STM inhibits HCC, we chose poorly differentiated and more malignant HCC cell line SK-Hep-1 to establish in vivo HCC model. We found that STM significantly inhibited the growth of SK-Hep-1 derived tumor in nude mice (Fig. 1E, F, G). Collectively, these data demonstrate that STM inhibits HCC.

#### Microarray analysis of differential genes in HepG2 cells exposed to STM

To reveal the molecular mechanism by which STM inhibits the malignant phenotypes of HepG2 cells, we performed microarray analysis to screen differential genes in HepG2 cells exposed to STM. Compared to cells exposed to vehicle control, XXX genes were significantly upregulated and XXX genes were significantly downregulated in HepG2 cells exposed to STM (FC >2, P < 0.01, and AUC =1, Fig. 2A, Supplemental Table 1). Among the most significantly downregulated genes we selected 7 genes for PCR verification, and the results showed good correlation with microarray data (Fig. 2B).

## Pathway and biological process analysis of differentially expressed genes

Next we performed pathway analysis to identify significant pathways of differentially expressed gene (DEGs) in accordance with KEGG (Fig. 3A). We also classified the DEGs into different biological processes (Fig. 3B). Among the significant pathways of DEGs, PI3K-Akt signaling pathway showed the highest P



Figure 2. The differential expressed genes in HepG2 cells treated by STM. A. Differential expressed genes (DEGs) identified in STM treated HepG2 cells. The x-axis is log2 ratio of gene expression levels between two cells; the y-axis is p value based on -log10. The red dots represent the very significantly DEGs (P<0.01 and absolute value of log2(FC) < 1); the blue dots represent the other DEGs (P<0.01 and absolute value of log2(FC) < 1); the gray blue dots represent the transcripts whose expression levels did not reach statistical significance (P>0.01). B. PCR analysis of seven DEGs.

value, thus we further analyzed the expression changes of the DEGs involved in PI3K-Akt signaling pathway. The DEGs upregulated by STM were indicated by red while those downregulated by STM were indicated by green (Fig. 4).

## Prediction of targets of STM

To understand the mechanism underlying the anti-cancer activity of STM on HepG2 cells, we the targets of STM. Chemical predicted genomics-based virtual screening (CGBVS) could predict compound-protein interactions (CPIs) by using a support vector machine [8]. K Nearest Neighbor with K=3 (3NN) is another powerful program to predict and characterize drug-target associations [9]. We found that 47 targets of STM were predicted by CGBVS while 21 targets of STM were predicted by 3NN. Notably, 8 targets were predicted as the targets of STM by both programs (Fig. 5). Network analysis of these 8 targets and the most DEGs showed the potential functional interaction between them, which may then regulate the result gene (effector gene) to execute the activity of STM (Fig. 6).

#### **Regulation network of JUN and STAT3**

Among the 8 predicted targets of STM, JUN and STAT3 are two important transcription factors known

to regulate the expression of a wide variety of downstream target genes. Therefore, we performed network analysis of the DEGs whose expression is regulated by JUN and STAT3. We found that a large range of DEGs with different expression changes induced by STM could be regulated by JUN and STAT3 (Fig. 7).

#### Discussion

STM is a secoiridoid glycoside mainly isolated from Enicostemma species and has demonstrated a wide variety of biological activities such as anti-inflammatory, anti-pyretic, anti-microbial, anti-malarial, and anti-diabetic activity [10-16]. However, the anti-tumor activity of STM has seldom been reported. In this study, we first evaluated the anti-tumor activity of STM on HCC cells. Next we explored gene expression profiling of HepG2 cells treated by STM and performed bioinformatics analysis to elucidate the molecular mechanism and identify potential targets of STM to inhibit HCC.

Cell proliferation is known to be crucially regulated by cell cycle control. Chemoradiotherapy agents mainly inhibit tumor growth via the regulation of the expression and function of cell cycle regulatory proteins [17,18]. Notably, a previous study reported that STM inhibited the proliferation and increased the apoptosis of fibroblast-like synoviocytes and this was related to the upregulation of caspase 3, an important effector of apoptosis, by STM in fibroblast-like synoviocytes [19]. These findings are consistent with our data that STM inhibited the proliferation and increased the apoptosis of HepG2 cells. However, we did not find caspases in the list of DEGs. Instead, we found that BCL2L13 (BCL2-like 13), an apoptosis facilitator, was upregulated by STM in HepG2 cells. In addition, we found that cyclin-dependent kinase inhibitor 1A (CDKN1A, also known as p21), cyclin-dependent kinase inhibitor 1B (CDKN1B, also known as p27), cyclin-dependent kinase inhibitor 1C (CDKN1C, also known as p57) were all upregulated by STM in HepG2 cells (please see supplemental table 1). Collectively, these results suggest that STM inhibits HCC cell growth via the blockage of cell cycle progression and the induction of cell apoptosis.

MMPs are crucially involved in the degradation of extracellular matrix and facilitate tumor invasion and metastasis. STM has been shown to downregulate the expression of MMP1 and MMP3 in fibroblast-like synoviocytes [19]. In this study, we did not find MMP1 and MMP3 in the list of DEGs. Instead, we found that MMP2 expression was downregulated by STM in HepG2 cells. Thus we speculate that STM may mainly downregulate MMP2 to inhibit the invasion ability of HepG2 cells. The differential regulation of MMPs by STM in different cell types may be due to cell type specificity. Further studies are needed to reveal the anti-metastasis mechanism of STM in HCC.

Using two different programs CGBVS and 3NN, we identified 8 targets of STM that were predicted by both programs. Among the 8 predicted targets of STM, JUN and STAT3 attracted our attention because they regulate the expression of a wide variety of downstream target genes which could modulate cell proliferation, apoptosis and invasion [20-22]. CDKN1A expression is known to be repressed by JUN but its expression was significantly upregulated by STM in HepG2 cells (please see CDKN1A in bright red in Fig. 7A). On the other hand, MMP2 expression is known to be activated by JUN but its expression



Figure 3. Pathways and biological processes significantly enriched by differential expressed genes. A. KEGG pathways enriched DEGs. B. Biological Processes enriched DEGs. The line width indicates the enrichment percentage. The dotted line in the box indicates the significance threshold (P value = 0.05).

was significantly downregulated by STM in HepG2 cells (please see MMP2 in bright green in Fig. 7A). These results indicate that JUN is a target of STM and its activity is inhibited by STM. Similarly, MMP2 expression is known to be activated by STAT3 but its expression was significantly downregulated by STM in HepG2 cells (please see MMP2 in bright green in Fig. 7B). In addition, STM attenuated inflammation via inhibiting JAK2/STAT3 signaling in adjuvant induced arthritis [23]. STAT3 is known to be activated to regulate cell survival and apoptosis in response to stress [24,25]. Therefore, STAT3 is another target of STM and its activity is inhibited by STM. Moreover, STW was shown to attenuate carbon tetrachloride-induced liver fibrosis via the inhibition of TGFB1/SMAD pathway [26]. In addition, a recent study revealed a common pathway that regulated

both MMPs and Smad4 in vulvar squamous cell cancer [27]. While we did not find SMAD, TGFß1 was shown in Fig. 7B. Next we will perform functional studies to verify the predicted targets of STM, which will explain how STM exerts anti-tumor efficacy in HCC.

In conclusion, STM treatment led to significant changes in the expression of a variety of genes that modulate cell survival, cell cycle progression, apoptosis, and invasion. Moreover, most of these genes can be clustered into pathway networks such as PI3K, JUN, STAT3, which are predicted targets of STM. Further confirmation of these targets will reveal the anti-tumor mechanisms of STM and pave the way for the development of STM as a novel agent for cancer prevention and treatment.



Figure 4. Gene expression profiles of the PI3K-AKT signaling pathway in treated HepG2 cell. The red and green colors represent the log2(FC) of the corresponding genes.

A	CGBVS 3NN 47 8 21				
В	Gene name	Protein name	Uniprot	3NN score	RF_Prob
	PRKCH	Protein kinase C eta	P24723	0.46	0.55
	JUN	Proto-oncogene c-JUN	P05412	0.43	0.59
	PTPN1	Protein-tyrosine phosphatase 1B	P18031	0.53	0.62
	STAT3	Signal transducer and activator of transcription 3	P40763	0.44	0.55
	PRKCE	Protein kinase C epsilon	Q02156	0.48	0.55
	ACP1	Low molecular weight phosphotyrosine protein phosphatase	P24666	0.46	0.57
	TLR9	Toll-like receptor 9	Q9NR96	0.40	0.60
	CDC25B	Dual specificity phosphatase Cdc25B	P30305	0.53	0.55

Figure 5. Venn diagram of the predicted targets by two different methods and the consistent targets. A. Venn diagram of the predicted targets by CGBVS and 3NN. B. Eight targets predicted by the two methods.



Figure 6. Network analysis of the consistent targets and the most differential expressed genes. The network shows the predicted functional interaction between the consistent targets (pink circle) and most DEGs (purple circle) according to the Genemania database. The gray nodes represents the result genes involved in the functional interaction between the pink and purple circles. Each colored line represents a different interaction, and the color line width indicates the confidence of the interactions.



Figure 7. Regulation analysis of transcription factors JUN and STAT3. A. Regulatory model of JUN in the treated cell, including the regulation types of activation, repression, and unknown. B. Regulatory model of STAT3 in the treated cell, including the regulation types of activation, repression, and unknown. The gradient color from red to green is expressed as the logFC value of each gene.

## **Supplementary Material**

Supplementary table. http://www.jcancer.org/v10p6526s1.xlsx

#### Acknowledgements

This work was supported by the grants from the National Natural Science Foundation of China (No. 81660399 and 81860423), the Innovative Research Team Project of Yunnan Province (No. 2015HC033), the Yunnan Provincial Academician Workstation of Xiaoping Chen (No. 2017IC018), the Breeding Program for Major Scientific and Technological Achievements of Kunming Medical University (No. CGYP201607), the Medical Leading Talent Project of Yunnan Province (No. L201622), and Yunnan Provincial Clinical Center of Hepato-biliarypancreatic Diseases to L. Wang; Scientific Research Foundation of the Department of Education of Yunnan Province (No. 2019J1253) to H.R. Tang; and the Leading Academic and Technical Young and Mid-aged Program of Kunming Medical University (No. 60118260108) to Y. Ke.

## **Competing Interests**

The authors have declared that no competing interest exists.

## References

- Ferlay J, Colombet M, Soerjomataram I, et al. Estimating the global cancer incidence and mortality in 2018: GLOBOCAN sources and methods. Int J Cancer. 2018; 144:1941-1953.
- [2] Wei D, Wang L, Chen Y, et al. Yangyin Fuzheng Decoction enhances anti-tumor efficacy of cisplatin on lung cancer. J Cancer. 2018; 9: 1568-1574.
- [3] Villa-Hernández JM, García-Ocón B, Sierra-Palacios EC, et al. Molecular biology techniques as new alternatives for medicinal plant identification. Phyton, Int J Exp Botany. 2018; 87: 72-78.
- [4] Zhong J, Deng L, Jiang Y, et al. Gene expression profiling of HepG2 cells after treatment with black tea polyphenols. Biocell. 2018; 42: 99-104.
- [5] Jaishree V, Badami S. Antioxidant and hepatoprotective effect of swertiamarin from Enicostemma axillare against D-galactosamine induced acute liver damage in rats. J Ethnopharmacol. 2010; 130: 103-106.
- [6] Montes FQ, Vázquez-Hernández A, Fenton-Navarro B. Active compounds of medicinal plants, mechanism for antioxidant and beneficial effects. Phyton, Int J Exp Botany. 2019; 88: 1-10.
- [7] Zhang L, Cheng Y, Du X, et al. Swertianlarin, an Herbal Agent Derived from Swertia mussotii Franch, Attenuates Liver Injury, Inflammation, and Cholestasis in Common Bile Duct-Ligated Rats. Evid Based Complement Alternat Med. 2015; 2015: 948376.
- [8] Hamanaka M, Taneishi K, Iwata H, et al. CGBVS-DNN: Prediction of Compound-protein Interactions Based on Deep Learning. Mol Inform. 2017; 36: 1-2.
- [9] Liu X, Xu Y, Li S, et al. In Silico target fishing: addressing a "Big Data" problem by ligand-based similarity rankings with data fusion. J Cheminform. 2014; 6: 33.
- [10] Vaidva H, Prajapati A, Rajani M, et al. Beneficial Effects of Swertiamarin on Dyslipidaemia in Streptozotocininduced Type 2 Diabetic Rats. Phytother Res. 2012; 26: 1259-1261.
- [11] Sonawane RD, Vishwakarm SL, Lakshmi S, et al. Amelioration of STZ-induced type 1 diabetic nephropathy by aqueous extract of Enicostemma littorale Blume and swertiamarin in rats. Mol Cell Biochem. 2010; 340: 1-6.
- [12] Vaidya H, Rajani M, Sudarsanam V, et al. Swertiamarin: a lead from Enicostemma littorale Blume. for antihyperlipidaemic effect. Eur J Pharmacol. 2009; 617: 108-112.
- [13] Vaidya H, Giri S, Jain M, et al. Decrease in serum matrix metalloproteinase-9 and matrix metalloproteinase-3 levels in Zucker fa/fa obese rats after treatment with swertiamarin. Exp Clin Cardiol. 2012; 17: 12-16.

- [14] Vaijanathappa J, Badami S. Antiedematogenic and free radical scavenging activity of swertiamarin isolated from Enicostemma axillare. Planta Med. 2009; 75: 12-17.
- [15] Anjum L, Ansari A, Yadav A, et al. Quantitative determination of Swertiamarin in Swertia chirayita by HPTLC. Int Eng Res Gen Sci. 2014; 2: 85– 92.
- [16] Leong XY, Thanikachalam PV, Pandey M, et al. A systematic review of the protective role of swertiamarin in cardiac and metabolic diseases. Biomed Pharmacother. 2016; 84: 1051-1060.
- [17] Cai Z, Liu Q. Cell Cycle Regulation in Treatment of Breast Cancer. Adv Exp Med Biol. 2017; 1026: 251-270.
- [18] Soltanian S, Riahirad H, Pabarja A, et al. Kaempferol and docetaxel diminish side population and down-regulate some cancer stem cell markers in breast cancer cell line MCF-7. Biocell. 2017; 41: 33-40.
- [19] Saravanan S, Islam VI, Thirugnanasambantham K, et al. Swertiamarin ameliorates inflammation and osteoclastogenesis intermediates in IL-1b induced rat fibroblast-like synoviocytes. Inflamm Res. 2014; 63: 451–462.
- [20] Liu J, Fei D, Xing J, et al. MicroRNA-29a inhibits proliferation and induces apoptosis in rheumatoid arthritis fibroblast-like synoviocytes by repressing STAT3. Biomed Pharmacother. 2017; 96: 173-181.
- [21] Ke Y, Bao T, Zhou Q, et al. Discs large homolog 5 decreases formation and function of invadopodia in human hepatocellular carcinoma via girdin and tks5. Int J Cancer. 2017; 141: 364-376.
- [22] Ke Y, Bao T, Wu X, et al. Scutellarin suppresses migration and invasion of human hepatocellular carcinoma by inhibiting the stat3/girdin/akt activity. Biochem Biophys Res Commun. 2017; 483: 509-515.
- [23] Saravanan S, Islam VI, Babu NP, et al. Swertiamarin attenuates inflammation mediators via modulating NF-κB/I κB and JAK2/STAT3 transcription factors in adjuvant induced arthritis. Eur J Pharm Sci. 2014; 56: 70-86.
- [24] Ye H, Wang W, Ding Y, et al. Spontaneous running wheel improves neuroprotection efficacy of ischemic postconditioning in mice following ischemia/reperfusion injury. Biocell. 2018; 42: 79-85.
- [25] Wang C, Zhou X, Xu H, et al. Niclosamide Inhibits Cell Growth and Enhances Drug Sensitivity of Hepatocellular Carcinoma Cells via STAT3 Signaling Pathway. J Cancer. 2018; 9: 4150-4155.
- [26] Chen K, Wu T, Zhang RY, et al. Effects of swertiamarin on TGF-ß1/Smad signaling pathway in rats with carbon tetrachloride-induced liver fibrosis. Int J Clin Exp Med. 2017; 10: 2316-2325.
- [27] Yang X, Wu X. MiR-182-5p promotes proliferation and invasion by regulating Smad4 in vulvar squamous cell cancer. Eur J Gynaecol Oncol. 2018; 39: 604-608.