

Differential Expression of Circular RNAs in Glioblastoma Multiforme and Its Correlation with Prognosis¹



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Abstract

OBJECTIVE: The present study aimed to explore the expression profiles of circular RNAs (circRNAs) in glioblastoma multiforme (GBM) in an attempt to identify potential core genes in the pathogenesis of this tumor. **METHODS:** Differentially expressed circRNAs were screened between tumor tissues from five GBM patients and five normal brain samples using Illumina HiSeq. Bioinformatics analysis was used to analyze their potential function. CircBRAF was further detected in different WHO grades glioma tissues and normal brain tissues. Kaplan-Meier curves and multivariate Cox's analysis were used to analyze the association between circBRAF expression level and prognosis of glioma patients. **RESULTS:** A total of 1411 differentially expressed circRNAs were identified in GBM patients including 206 upregulated circRNAs and 1205 downregulated circRNAs. Differential expression of circRNAs was closely associated with the biological process and molecular function. The downregulated circRNAs were mainly associated with ErbB and Neurotrophin signaling pathways. Moreover, the expression level of circBRAF in normal brain tissues was significantly higher than that in glioma tissues ($P < .001$). CircBRAF was significantly lower in glioma patients with high pathological grade (WHO III & IV) than those with low grade (WHO I & II) ($P < .001$). Cox analysis revealed that high circBRAF expression was an independent biomarker for predicting good progression-free survival and overall survival in glioma patients (HR = 0.413, 95% CI 0.201-0.849; HR = 0.299, 95% CI 0.135-0.661; respectively). **CONCLUSION:** The present study identified a profile of dysregulated circRNAs in GBM. Bioinformatics analysis showed that dysregulated circRNAs might be associated with tumorigenesis and development of GBM. In addition, circBRAF could serve as a biomarker for predicting pathological grade and prognosis in glioma patients.

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Introduction

Glioma is the most common primary intraparenchymal tumor of the central nervous system (CNS). It can be classified into four grades as WHO I to IV according to the WHO criteria 2007 [1], among which glioblastoma multiforme (GBM) belonging to WHO grade IV is the most aggressive form with rapid recurrence. Currently, surgery, postoperative radiotherapy, and temozolomide-based chemotherapy are the mainstay treatments for GBM. However, the highly proliferative, aggressive, and invasive nature of GBM contributes to dismal prognosis, with a mean survival of only 14.6 months [2]. To better understand and improve the therapeutic efficacy, many studies have focused on the genome mutation and transcriptome instability in GBM [3–5].

CircRNAs are a group of non-protein-coding RNAs identified from some transcribed genes about 20 years ago [6–8]. Recent studies have demonstrated their widespread distribution and substantial presence by using high-throughput sequencing and novel computational approaches [9–11], suggesting that circRNAs are a family of naturally occurring endogenous ncRNAs with single-stranded covalently closed circular molecules with neither 5'-3' polarity nor a polyadenylated tail [11]. CircRNAs have been characterized in several tumors and cell lines [12–17]. Cumulative studies have indicated that some circRNAs play an important role in the pathogenesis and development of cancer. Circ_001569 can promote cell proliferation and invasion of colorectal cancer by sponging miR-145 and upregulating miR-145 functional targets E2F5, BAG4, and FMNL2 [18]. Another study also found that silencing of circHIPK3 could inhibit cell proliferation of several tumor cell lines such as Huh7, HCT-116, and HeLa [14]. A recent study found that circZNF292 was positively associated with human glioma tube formation [19].

Despite the findings of the existing studies suggesting that circRNAs play an important role in the development of many tumors, the expression profile and potential function of circRNAs in glioma remain unclear. The aim of the present study was to explore the expression profiles of circRNAs in GBM in an attempt to identify potential core genes in the pathogenesis

of this CNS malignant tumor and to furthermore determine the association between circRNAs and the prognosis of glioma patients.

Material and Methods

Acquisition of the Clinical Specimens

Five GBM and five normal brain specimens were obtained from patients who underwent surgery at Changzheng Hospital (Shanghai, China). GBM was diagnosed according to the 2007 WHO classification of tumors of the CNS [1]. Another 68 Chinese glioma patients (24 female and 44 male; age: 50.18 ± 15.7 years) underwent first surgical resection in Changzheng Hospital between March 2012 and October 2015. All glioma patients did not receive chemotherapy or radiotherapy before surgery. The clinical, surgical, imaging, and pathologic records of the patients were reviewed retrospectively. The pathologic diagnosis was confirmed by two experienced pathologists independently, and histopathological grade was determined according to the 2007 WHO Classification of Tumors of the Central Nervous System [1]. Tumor recurrence and progression were defined as clinical and/or imaging progression on follow-up magnetic resonance imaging findings. All the surviving patients were followed up for at least 1 year on the outpatient basis at 3-month intervals for the first 6 months, then at 6-month intervals for the next 2 years, and annually for life thereafter. The study was performed in accordance with the 1964 Helsinki Declaration and subsequent amendments or comparable ethical standards. The study protocol was approved by the Ethic Committee of Changzheng Hospital affiliated to the Second Military Medical University (Shanghai, China), and informed consent was obtained from the surviving patients or family members of those who had died.

Sequencing Library Preparation and circRNA Sequencing

Total RNA was isolated from the specimens using RNAiso Plus reagent (TaKaRa, Japan) according to the manufacturer's protocol. The NanoDrop ND-1000 spectrophotometer (Agilent Inc., USA) was used to measure RNA quantification and quality. RNA integrity and gDNA contamination were tested by denaturing agarose gel electrophoresis. Sequencing library was determined by Agilent 2100 Bioanalyzer using the Agilent DNA 1000 chip kit (Agilent, part# 5067-1504, USA). Total RNA from each sample was used to prepare the circRNA sequencing library via the following steps: 1) 5 µg total RNA was pretreated to enrich circRNA using circRNA Enrichment Kit (Cloud-seqInc, USA); 2) RNA libraries were constructed by using pretreated RNAs with TruSeq Stranded Total RNA Library Prep Kit (Illumina, USA) according to the manufacturer's instructions; 3) the libraries were controlled qualitatively and quantitatively by using the BioAnalyzer 2100 system (Agilent Technologies, Inc., USA),

Table 1. Primers and DNA Sequences Used in This Study

circRNA	Primer Sequence (5'-3')
hsa:circ_0000199	F:TCATTGCTTTCAGGGCTCTT R:ATAGAAACGTGTGCGGTCCCT
hsa:circ_0002968	F:GGAGCTCATGGATGCAAATC R:TTGTGTGCACGCTTGCTTCT
hsa:circ_0005603	F:ATCAGTGGACAGCTTGTGGA R:TCCTGATGACTCAATGCTGTG
hsa:circ_0008345	F:GGCCAGAAGAAGGAATCAG R:GCAATTTCTGGTCCACTTG
chr4:103,501,692–103,504,114+	F:TGGAGTCTGGGAAGGATTTG R:ATTTCTCCCTCCAGTCAC
hsa:circ_0003586	F:CCTTAACTATCCAAAAGCCAAA R:TCAATATTTCCCGCAACCAC
hsa:circ_0004872	F:TGCAGATCCAGACCATGATCACA R:TTCTCATGTCTGAAGCGCAGT
hsa:circ_0006411	F:TCTCGCCTCCACACCAAAA R:TCCACACACGAGCTTCCAC
hsa:circ_0001566	F:CTTTGTGGTATTAACATCTGC R:TCACATTTACTGTGCTGCTCA
circBRAF	F:CTCCAGCTTGATCACCAT R:TCTTCATCTGCTGGTCCGAA
chr19:5,761,476–5,761,630+	F:TCACAGTTCTCAGGCTTAACA R:ATACACCATCCAGCGCCTTC
chr18:107,887–108,499-	F:GCCCCTTTAGGCAGAGCTTA R:CTACAGGGGGCTTGTGACAT
BRAF	F:GAAGACCTCACAGTAAAAATAGGTGA R:CCACAAAATGGATCCAGACA

Table 2. Reads Statistics

Sample	Raw Reads	Mapped Reads	circRNA Number
N1	63,504,910	54,253,992	6928
N2	95,430,420	80,774,288	9112
N3	97,418,358	80,320,270	7839
N4	116,474,654	77,690,582	7418
N5	73,894,064	48,289,650	7496
T1	109,894,058	92,221,972	23,113
T2	40,432,568	24,233,908	3139
T3	66,533,768	53,711,986	9178
T4	60,880,950	43,383,662	2733
T5	65,515,164	41,651,528	4507

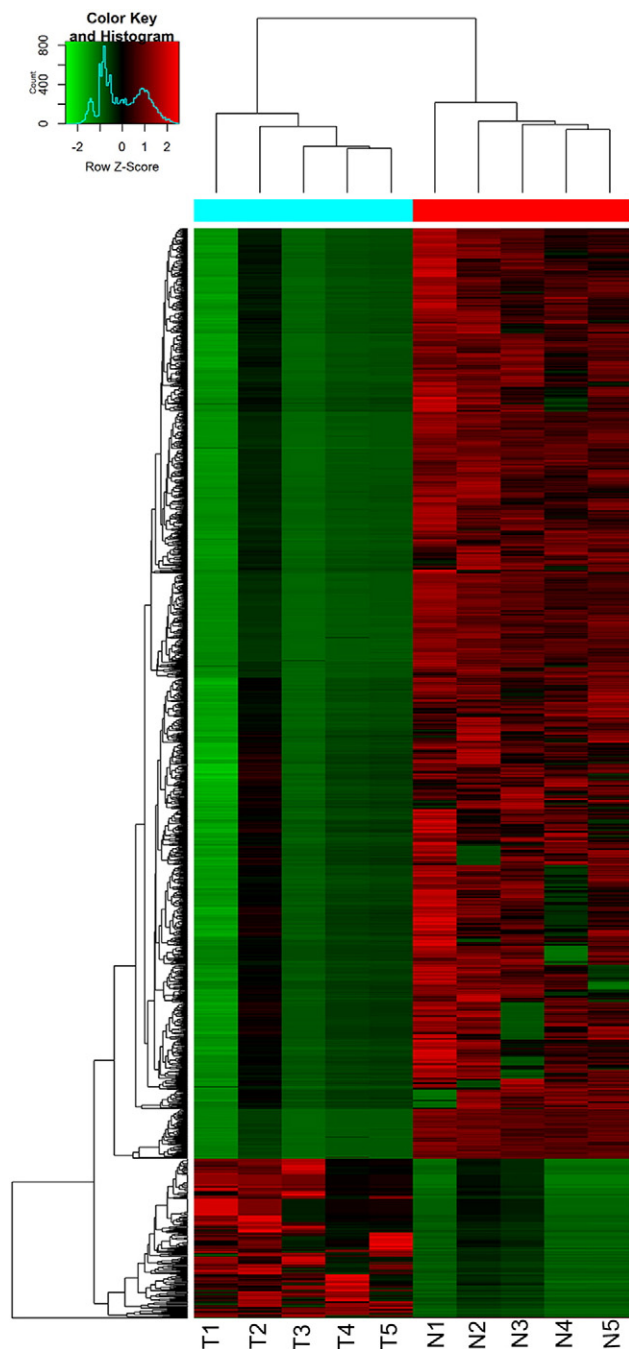


Figure 1. Hierarchical cluster analysis of circRNAs; unique circRNA expression profile of the 10 samples in red (upregulated expression) or green (downregulated expression) scale. The dendrogram shows the relationships among the expression levels of samples.

denatured as single-stranded DNA molecules, captured on Illumina flow cells, amplified *in situ* as clusters, and finally sequenced for 150 cycles on Illumina HiSeq Sequencer according to the manufacturer's instructions.

RNA-seq Data Statistics and RNA Labeling

Compared with the classical linear transcripts, circRNAs were covalently closed. Circular RNA molecules were spliced at the canonical splice site. The back-splicing reads indirectly represented the expression quality of circRNAs. After image recognition and base recognition, raw sequencing reads were acquired from Illumina HiSeq

Sequencer. The total number of reads in the samples was obtained for standardization and log₂ conversion, and the number of standardized reads was compared between the samples. The raw reads and the human reference genome (hg19) were compared using STAR software [20], and the Ensemble (V70) GTF file was used as the guide. CircRNAs were detected using the DCC software [21]. Altogether 59,386 circRNAs candidates were detected in these samples. CircBase database and circ2Trait disease database were used to label the circRNAs.

Identification of Differentially Expressed circRNAs

After normalization of the raw reads, circRNAs belonging to at least two samples were chosen for further data analysis. Using the normalized number of reads, significantly differentially expressed circRNAs were identified through Volcano Plot filtering. Hierarchical clustering was performed using Heatmap2 in software R, and circRNAs with a fold change more than 2.0, *P* value less than .05, and false discovery rate (FDR) less than 0.05 were filtered.

Quantitative Real-Time Polymerase Chain Reaction (qRT-PCR)

To validate the accuracy of the circRNAs-seq data, qRT-PCR was performed as a confirmatory method. cDNA of the tumor and normal samples was synthesized by reverse transcription using a Primescript RT reagent kit with gRNA Eraser according to manufacturer's protocols (TaKaRa, Japan). The differentially expressed circRNAs were measured by qRT-PCR using SYBR premix ExTaq (TaKaRa, Japan) on Applied Biosystems StepOnePlus Real-Time PCR system. The primers of randomly selected 12-pair outward-facing circRNAs were shown in Table 1. The qRT-PCR condition was set at an initial denaturation step of 10 minutes at 95°C and 95°C for 10 seconds, 60°C for 60 seconds, 95°C for 10 seconds for a total of 40 cycles, with a final step heating slowly from 60°C to 99°C. The relative expression of each circRNA was calculated relative to GAPDH.

Bioinformatics Analysis of Differentially Expressed circRNAs

CircRNAs were generated from intron-derived RNAs and back-spliced exons. The parent linear mRNAs of differentially expressed circRNAs were studied with Gene Ontology (GO) (<http://www.geneontology.org>) and Kyoto Encyclopedia of Genes and Genomes (KEGG) (<http://www.genome.jp/kegg>) databases. Based on GO and KEGG pathway analysis, differentially expressed circRNAs were annotated and their functions were presumed.

Interactions between circRNAs and miRNAs

Interactions between differentially expressed circRNAs and miRNAs were investigated by using miRanda and Target Scan databases for predicting target miRNAs, whereby circRNAs that may function as miRNA sponges could be inferred. The network between circRNAs and miRNAs was also constructed by Cytoscape based on the binding sites of the differentially expressed circRNAs and miRNAs.

Validation of circBRAF Expression in Glioma Patients

GO and pathway analysis showed that circBRAF located in chr7:140,476,711-140,494,267 was a potential core gene that might be related to tumorigenesis of glioma. The relative expression of circBRAF and BRAF mRNA was validated by qRT-PCR in the five GBM and five normal brain samples. The relative expression of circBRAF was validated in 68 glioma species with different grades and 13 normal brain species by qRT-PCR as described above. Also, the

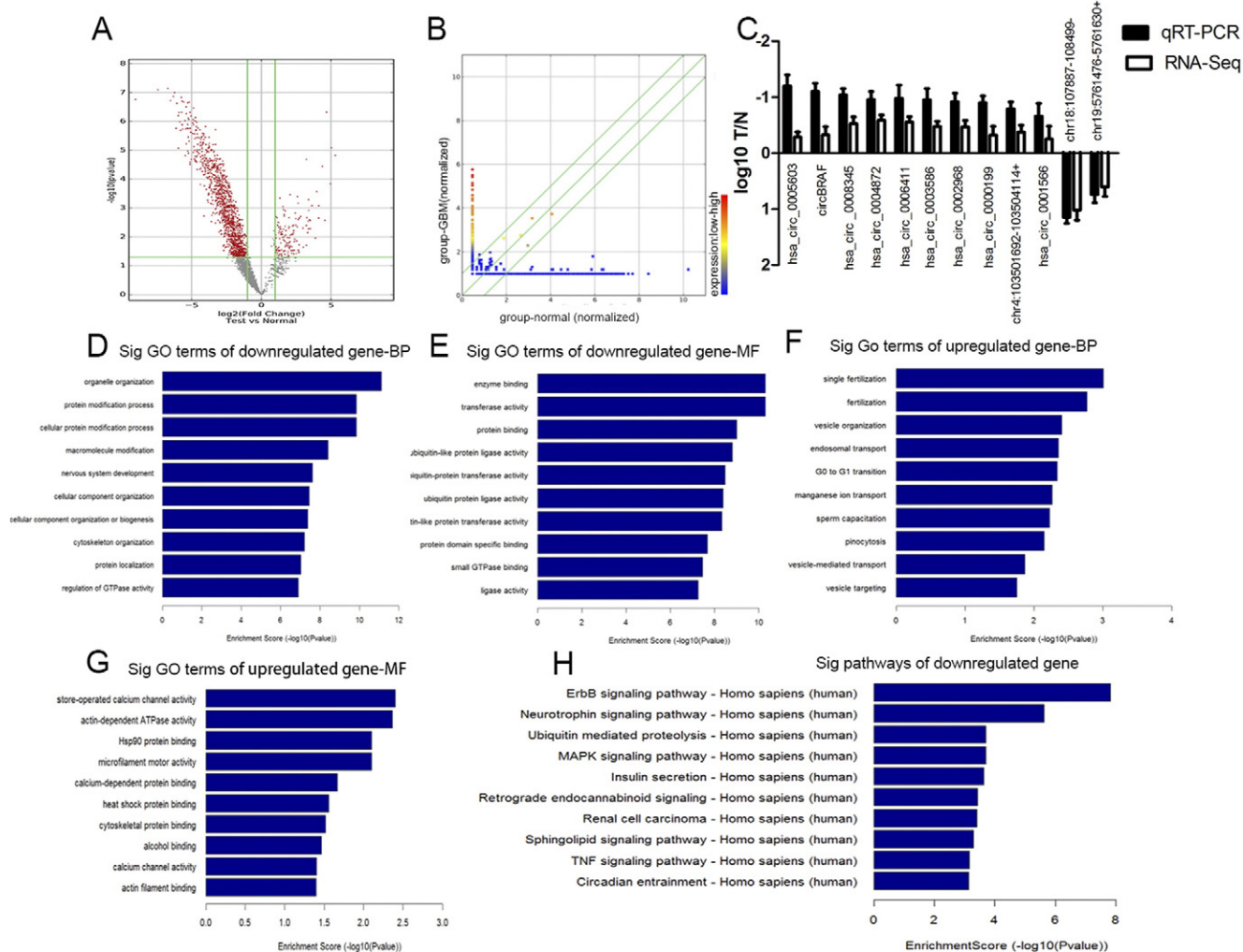


Figure 2. Volcano plot filtering out the differentially expressed circRNA (red frame represented differentially expressed circRNA with fold change > 2, $P < .05$, $FDR < 0.05$) (A); scatter plot (red dots and blue dots represented highly abundant circRNAs and low-abundance circRNAs) (B); good correlation of results from circRNA-seq with RT-PCR (12 circRNAs) (C); GO annotation analysis of downregulated circRNAs with top 10 enrichment score (D) and molecular functions (E); GO annotation analysis of upregulated circRNAs with top 10 enrichment score (F) and molecular functions (G); KEGG pathway enrichment analysis of downregulated circRNAs with top 10 enrichment score (H).

association of circBRAF with prognosis of glioma patients was further evaluated.

Statistical Analysis

Chi-square test or Fisher's exact test was used to analyze relevant factors. The Kaplan-Meier method was used to estimate the survival probability. The difference in survival probability between groups was analyzed by log-rank tests. Multivariate analysis of prognosis factors was performed using a Cox regression analysis. For the stability of the equation, the histology grade was not included in the multivariate analysis. All statistical analyses were performed in SPSS 20.0 (SPSS, Inc., Chicago, IL). $P < .05$ was considered statistically significant.

Results

Differential Expression of circRNAs in Tumor Tissues from Glioma Patients

CircRNAs were sequenced in five GBM and five normal brain samples. The circRNA-seq reads of each sample are shown in Table 2.

There were 12,048 circRNAs overlapped when compared to Rybak-Wolf's research on human brain [22]. Hierarchical clustering analysis clearly screened the differentially expressed circRNAs between these two groups (Figure 1). As shown in Figure 2A, a total of 1411 differentially expressed circRNAs (fold change ≥ 2.0 , $P < .05$, $FDR < 0.05$) were identified in glioma patients by Volcano Plot, including 206 upregulated circRNAs and 1205 downregulated circRNAs. Scatter plot analysis made the variation of circRNAs expressions of two groups be visualized in these plots (Figure 2B).

Validation of the Accuracy of circRNA-seq Data by qRT-PCR

To verify the differential expression of the candidate circRNA, qRT-PCR was conducted in five GBM and five normal samples. We selected 12 circRNAs (hsa:circ_0000199, hsa:circ_0002968, hsa:circ_0005603, hsa:circ_0008345, chr4:103,501,692–103,504,114+, hsa:circ_0003586, hsa:circ_0004872, hsa:circ_0006411, hsa:circ_0001566, chr19:5,761,476–5,761,630+, chr18:107,887–108,499–, circBRAF) according to the results of RNA sequencing. As shown in

circRNA-miRNA network

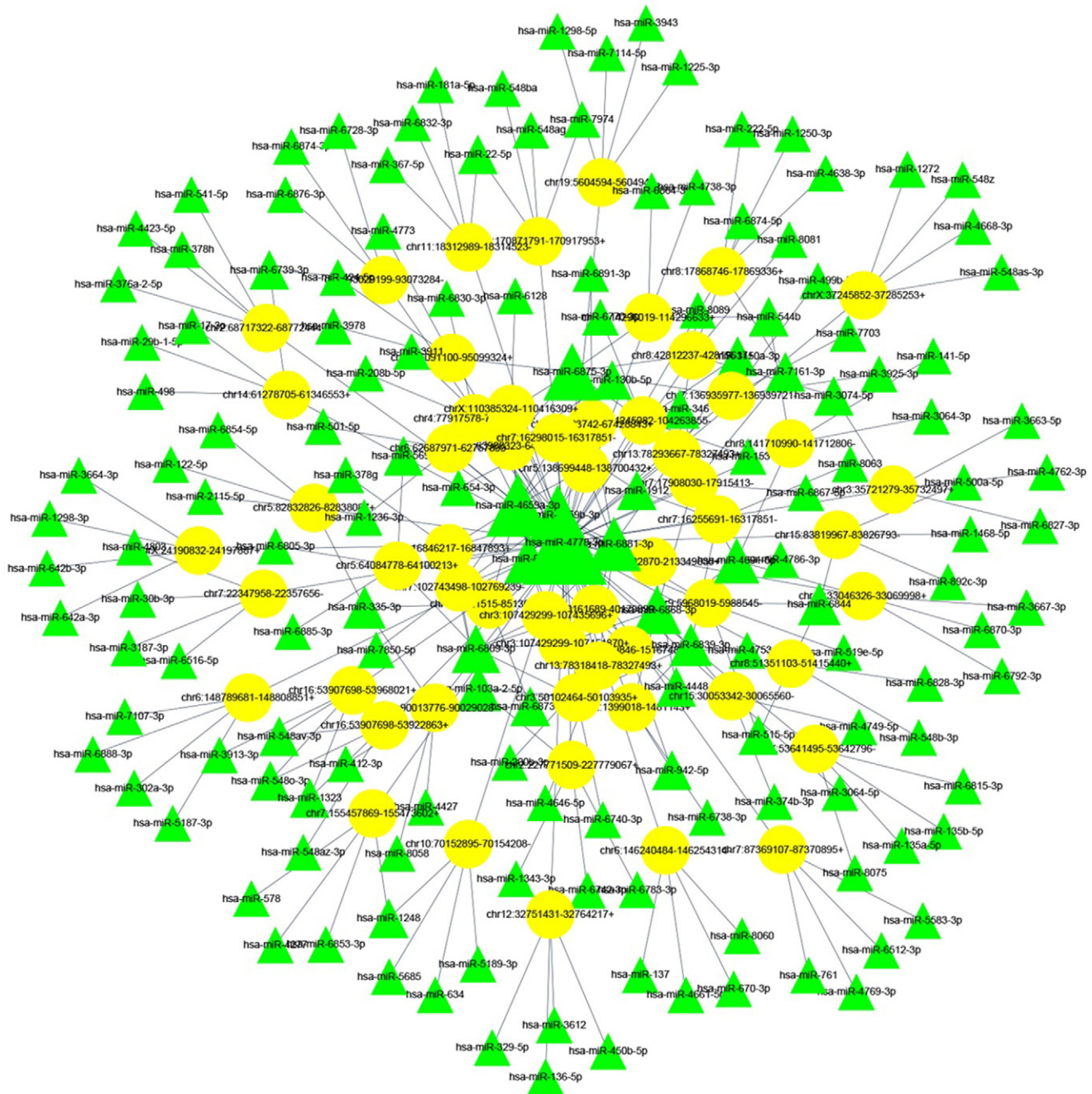


Figure 3. circRNA-miRNA network of 56 differentially expressed circRNAs and 154 predicted miRNAs.

Figure 2C, the results obtained from RT-PCR were consistent with RNA sequencing data.

GO Analysis and KEGG Pathway Analysis of Differentially Expressed circRNA

GO analysis showed that down-regulated circRNAs were in favor of the enzyme binding, transference activity, protein binding, organelle organization, and protein modification processes (Figure 2, D and E). As for upregulated circRNAs, the closely related GO terms were fertilization, G0 to G1 transition, manganese ion transport, sperm capacitation, store-operated calcium channel activity, and actin-dependent ATPase

activity (Figure 2, F and G). KEGG pathway analysis showed no enrichment pathway in the up-regulated circRNAs, but 70 pathways were identified to be associated with downregulated circRNAs. And ErbB signaling pathway and Neurotrophin signaling pathway were the top enrichment pathways in downregulated circRNAs (Figure 2H).

Identification of circRNA-miRNA Network

Bioinformatics analysis showed that some circRNAs functioned as miRNA sponges in mammal cells. As mentioned above, miRanda and Target Scan databases were used to predict miRNAs binding sites in circRNAs. Based on the differentially expressed circRNAs and predicted miRNA binding sites, 56

Table 3. Population Characteristics

Age	
Min-max	9-79
Mean (years \pm SD)	50.18 \pm 15.7
Sex	
Male, <i>n</i> (%)	44 (64.7%)
Female, <i>n</i> (%)	24 (35.3%)
KPS score (preoperative)	
\leq 70, <i>n</i> (%)	7 (10.3%)
$>$ 70, <i>n</i> (%)	61 (89.7%)
No. of lesions	
Single, <i>n</i> (%)	43 (63.2%)
Multiple, <i>n</i> (%)	25 (36.8%)
Extent of surgery	
Gross total resection (GTR), <i>n</i> (%)	46 (67.6%)
Subtotal resection (STR), <i>n</i> (%)	22 (32.4%)
Histology grade (WHO grade)	
Low (I & II), <i>n</i> (%)	22 (32.4%)
High (III & IV), <i>n</i> (%)	46 (67.6%)
Adjuvant treatment	
Radiotherapy, <i>n</i> (%)	50 (73.5%)
Chemotherapy, <i>n</i> (%)	56 (82.4%)
Recurrence, <i>n</i> (%)	39 (57.4%)
Death during follow-up, <i>n</i> (%)	34 (50.0%)
Follow-up period (months)	
Range	1-54
Mean (months \pm SD)	20.60 \pm 14.66

differentially expressed circRNAs and 154 predicted miRNAs were used to construct the circRNA-miRNA network (Figure 3).

Expression Level of circBRAF in Glioma Patients

The demographic feature and clinical characteristics of 68 glioma patients were presented in Table 3. Forty-six cases were high-grade (WHO III, 12; WHO IV, 34), while 22 cases were low-grade glioma (WHO I, 6; WHO II, 16). The circRNA-seq data showed that circBRAF derived from *BRAF* gene exon 8-13 was significantly downregulated in tumor group. The distinct product of the expected size was amplified using outward-facing primers and confirmed by Sanger sequencing (Figure 4A). There were no significant differences between circBRAF and BRAF mRNA expression in normal brain. However, circBRAF was significantly downregulated in GBM tissues than BRAF mRNA as indicated by qRT-PCR analysis (Figure 4B). Then, circBRAF was detected in 13 normal brain tissue specimens and 68 glioma specimens of different grades. The results showed that the relative expression level of circBRAF in the glioma tissues was significantly lower than normal brain tissues, and the relative expression level of circBRAF in high-grade glioma specimens was significantly lower than low-grade glioma specimens ($P < .001$) (Figure 4C). The correlations between circBRAF expression and other prognostic factors were shown in Table 4. The Kaplan-Meier analysis revealed a survival difference between high-circBRAF expression group and low-circBRAF expression group, suggesting that glioma patients with high circBRAF expression had significantly better progression-free survival (PFS) and overall survival (OS) than those with low circBRAF expression ($P < .001$ for both; Figure 4, D and E). Univariate analysis showed that high circBRAF expression, younger age, longer duration of symptoms, high preoperative Karnofsky Performance Status (KPS), low histologic grades, and single lobe lesion were closely associated with the longer PFS and OS (Table 5). In addition, multivariate Cox analysis revealed that high circBRAF expression was an independent factor for good PFS and OS in glioma patients (HR = 0.413, 95% CI 0.201-0.849; HR = 0.299, 95% CI 0.135-0.661;

respectively) (Table 6). Taken together, the high expression of circBRAF represented a lower grade of glioma and a poorer prognosis.

Discussion

CircRNA, emerging as important regulator of tumorigenesis, has been found to be associated with the development of several different type of cancers. It has been established that dysregulation of circRNA could contribute to the initiation and promotion of tumors. To clarify the potential role of circRNA in pathogenesis of glioma, we identified circRNA expression signature in tumor tissue from GBM patients and conducted the function analysis. In the present study, a total of 1411 differentially expressed circRNAs were observed in tumor tissues from glioma patients including 206 upregulated circRNAs and 1205 downregulated circRNAs. GO analysis and KEGG pathway analysis revealed that the differentially expressed circRNAs participated in several biological processes and signaling pathways such as ErbB signaling pathway and Neurotrophin signaling pathway. Several studies have shown that BRAF plays an important role in the development and progression of cancers [23]. But few studies have reported circular transcriptomes produced from the BRAF gene. Interestingly, we found that circBRAF, one of the downregulated circRNAs, was significantly associated with pathological grade and prognosis of glioma patients.

The present study showed that there was a significant difference in circRNAs expression between the tumor and normal brain tissues and most circRNAs were downregulated in GBM tissues. In consistent with our results, several previous studies have reported that circRNAs were abundant in the brain [15,24]. A similar phenomenon was also observed in progressive laryngeal cancer [12], breast cancer, colorectal cancer, gastric cancer, hepatocellular carcinoma, and prostate adenocarcinoma [14,17]. CircRNAs were significantly downregulated in these tumors when compared with matched normal tissues. On the contrary, some previous reports found that circRNAs were significantly upregulated in bladder urothelial carcinoma and kidney clear cell carcinoma [14]. Although the accurate role of circRNAs in tumorigenesis and prognosis of tumors remained unclear, we speculated that the abnormal expression of circRNAs and its skewed distribution in tumors might be a widespread phenomenon.

Accumulating evidences showed that circRNAs played a key role in many biological processes, and some circRNAs are identified as RNA binding protein and mRNA magnet to directly translate protein [11,25]. GO analysis showed that downregulated circRNAs were significantly associated with protein binding. Previous studies have found that circRNAs could bind and sequester RNA-binding proteins to form large RNA-protein complexes; these RNA-protein complexes might regulate the pool of RNA-binding proteins [26]. GO analysis also showed that of the closely GO terms related to upregulated circRNAs included G0 to G1 transition, manganese ion transport, store-operated calcium channel activity. Previous studies have found these biological processes may be involved in the development of glioma. For example, a previous study has found that cell cycle G0/G1 arrest could significantly affect glioma cell proliferation [27]. Also, some researchers have reported that Ca²⁺ might also be an important positive regulator of tumorigenesis in GBM in processes involving quiescence, maintenance, proliferation, or migration [28]. Based on our results, we hypothesized that upregulated circRNAs may regulate the G0/G1 transition or participate in the tumorigenesis of glioma by modulating Ca²⁺ transport. These require investigators to confirm in the future research. KEGG pathway analysis showed that ErbB and

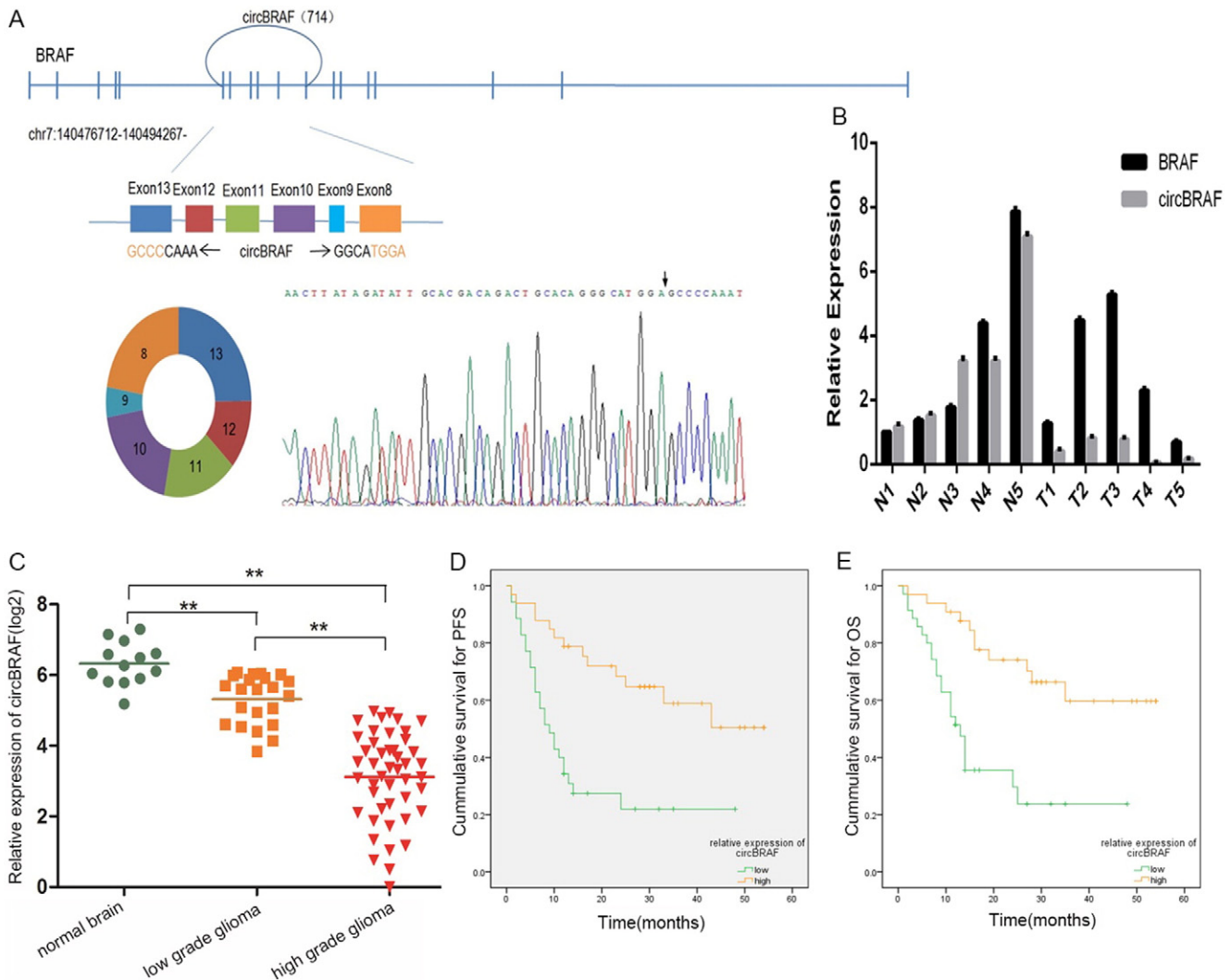


Figure 4. The genomic loci of circBRAF in BRAF gene and the expression of circBRAF was validated by qRT-PCR followed by Sanger sequencing (arrows represent divergent primers binding to the genome region of circBRAF) (A); the relative expression of circBRAF and BRAF mRNA in five normal brain and five GBM tissues (B); the relative expression of circBRAF in different grades glioma and normal brain tissues (C); Kaplan-Meier PFS curves of glioma patients in correlation with circBRAF expression (D); Kaplan-Meier OS curves of glioma patients in correlation with circBRAF expression (E); ***P* value < .01.

neurotrophin signaling pathways were the top enrichment pathways in downregulated circRNAs. Several studies have showed that the ErbB signaling pathway was evolutionarily conserved [29,30]. Blockade of ErbB signaling could inhibit the slow cycling JARID1B-positive cell population, which is critical for long-term maintenance of melanoma growth [30]. The ErbB signaling pathway could participate in metastasis of prostate cancer via regulation on epithelial-mesenchymal transition and stemness of prostate cancer cells [29]. That is to say, the ErbB signaling pathway may be closely related to proliferation, cell cycle, and differentiation of cancer cells. Currently, there are no studies focusing on the role of circRNAs in the ErbB pathway. The relationship between circRNAs and ErbB pathway needs to be further investigated. In most cases, neurotrophins have been shown to favor tumor development and progression. In glioma, neurotrophins (NGF, BDNF, and NT-3) and their receptors (TrkA, TrkB/TekC, and P75NTR) were detected in several lines of brain tumor-initiating cells. Furthermore, NGF, BDNF, and NT-3 were able to stimulate the proliferation of these tumor cells [31]. According to the results of KEGG analysis, downregulated

circRNAs may be involved in the pathogenesis of glioma through regulation of neurotrophins signaling pathway.

Recent functional research found that circRNAs could harbor miRNAs and function as miRNA sponges [32]. For instance, circCDR1a could harbor over 70 conserved binding sites for miR-7 and acts as a sponge for miR-7 to regulate the expression of human epidermal growth factor receptor, α -synuclein, and insulin receptor substrate-2 [32]. The murine circRNA circSry might harbor more than 10 miR-138 binding sites [13,32]. A recent study showed that circHIPK3 could sponge 9 miRNAs with 18 potential binding sites, and it also could directly bind to miR-124 and inhibit miR-124 activity [14]. To date, few studies have focused on the interaction of circRNAs and miRNAs in GBM. Here, we established a circRNA-miRNA network of GBM based on our RNA-seq data. The circRNA-miRNA network was able to help us further understand the role of circRNA in the tumorigenesis of glioma patients.

Previous studies found that BRAF played an important role in numerous human tumors. BRAF gene could participate in various pathways including ErbB signaling pathway, MAPK signaling

Table 4. Correlations between Relative Expression of circBRAF and Other Prognostic Factors

		Relative Expression of circBRAF (Number (% of Total))		P Value
		Low	High	
Sex	Male	25 (36.8)	19 (27.9)	.331
	Female	10 (14.7)	14 (20.6)	
Age (years)	<18	1 (1.5)	5 (7.4)	.011 *
	18-50	7 (10.3)	14 (20.6)	
	>50	27 (39.7)	14 (20.6)	
Duration of symptoms (months)	<12	29 (42.6)	24 (35.3)	.387
	>12	6 (8.8)	9 (13.2)	
KPS score	<80	30 (44.1)	31 (45.6)	.429
	>80	5 (7.4)	2 (2.9)	
Histologic grade	Low	1 (1.5)	21 (30.9)	<.001 **
	High	34 (50.0)	12 (17.6)	
Tumor size (cm)	<5	11 (16.4)	10 (14.9)	1.000
	>5	23 (34.3)	23 (34.3)	
No. of lesions	Single, n (%)	19 (27.9)	24 (35.3)	.137
	Multiple, n (%)	16 (23.5)	9 (13.2)	
Extent of surgery	Gross total resection (GTR)	24 (35.3)	22 (32.4)	1.000
	Subtotal resection (STR)	11 (16.2)	11 (16.2)	
Radiotherapy	No	10 (14.7)	8 (11.8)	.786
	Yes	25 (36.8)	25 (36.8)	
Chemotherapy	No	5 (7.4)	7 (10.3)	.534
	Yes	30 (44.1)	26 (38.2)	
Recurrence	No	9 (13.2)	20 (29.4)	.007 *
	Yes	26 (38.2)	13 (19.1)	
Survival	Yes	12 (17.6)	22 (32.4)	.015 *
	No	23 (33.8)	11 (16.2)	

Asterisks indicate statistical significance.

* P < .05.

** P < .01.

pathway, neurotrophin, and mTOR signaling pathway [30,33]. Our results showed that circBRAF derived from BRAF was significantly downregulated in GBM tumor tissue compared with normal brain tissue. Based on the extended experiment of the relative expression of circBRAF in different grades of glioma, we found that the relative expression level of circBRAF was lower in glioma patients with high pathologic grade than those with low grade. In addition, we evaluated the association of circBRAF expression with prognosis of glioma patients. Univariate and multivariate analyses showed that high circBRAF expression was significantly associated with good PFS and OS in glioma patients. Although the special function and mechanisms of circBRAF were unclear, we believe that circBRAF may serve as a biological biomarker for diagnosis or prognosis in glioma.

In conclusion, the present study identified a profile of dysregulated circRNAs in GBM. Bioinformatics analysis showed that dysregulated circRNAs might be associated with tumorigenesis and development of GBM and circBRAF could serve as a biomarker for predicting pathological grade and prognosis in glioma patients. The present study could widen the horizon of gene research in glioma and lay a preliminary foundation for further research of circRNAs in glioma.

Conflicts of Interest

None.

Authors' Contributions

Junle Zhu, Jingliang Ye, and Lei Zhang designed and carried out the experiments, and wrote the manuscript. Junle Zhu and Jingliang Ye

Table 5. Patient Characteristics and Univariate Analysis of Prognostic Factors Affecting PFS and OS.

Factors		PFS			OS		
		1-year Rate (%)	3-year Rate (%)	P Value	1-year Rate (%)	3-year Rate (%)	P Value
Sex	male/female	56.8 vs. 54.2	41.8 vs. 38.7	0.932	62.1 vs. 87.0	33.2 vs. 56.8	0.079
Age	>18/18-50/>50(years)	100 vs.76.2 vs.39.0	100 vs.47.5 vs.26.1	0.003 *	100 vs.85.7 vs.58.5	100 vs.57.0 vs.23.2	0.003 *
Duration of symptoms	<12/>12(months)	49.1 vs. 80.0	36.1 vs. 70.0	0.010 *	66.0 vs. 86.7	34.1 vs.68.9	0.041 *
KPS score	>70/<70	60.7 vs. 54.3	43.1 vs. 14.3	0.001 *	73.1 vs. 42.9	46.1 vs.0	0.001 *
Histology grade	low/high	95.5 vs. 37.0	83.9 vs. 18.8	<0.001 **	100 vs. 56.5	88.4 vs.17.7	<0.001 **
Tumor size (cm)	<5 cm/>5 cm	57.1 vs. 56.5	51.4 vs. 35.6	0.420	71.4 vs. 71.7	43.7 vs. 40.4	0.382
No. of lesions	single/multi	65.1 vs. 40.0	50.8 vs. 17.5	0.015 *	81.3 vs. 52.0	53.7 vs. 18.9	0.001 *
Extent of surgery	total/subtotal	56.5 vs. 54.5	45.4 vs. 28.6	0.598	76.1 vs. 58.7	48.0 vs.27.5	0.270
Radiotherapy	no/yes	55.6 vs. 56.0	37.5 vs. 38.8	0.668	66.7 vs. 71.9	47.1 vs. 37.3	0.829
Chemotherapy	no/yes	58.3 vs. 55.4	50.0 vs. 36.3	0.694	75.0 vs. 69.6	54.7 vs. 37.3	0.634
Relative expression of circBRAF	low/high	34.3 vs. 78.8	21.9 vs. 58.9	<0.001 **	51.4 vs. 74.1	23.7 vs. 59.7	<0.001 **

Asterisks indicates statistical significance.

* P < 0.05.

** P<0.01.

Table 6. Multivariate Analysis of Prognostic Factors for PFS and OS

Factors	PFS			OS		
	HR	95% CI	P Value	HR	95% CI	P Value
Age	2.203	1.143-4.244	.018 *	2.382	1.157-4.905	.019 *
Duration of symptoms	–	–	.006 *	–	–	.82
KPS score	2.623	1.060-6.490	.037 *	–	–	.092
No. of lesions	–	–	.078	2.878	1.396-5.936	.004 *
Relative expression of circBRAF	0.413	0.201-0.849	.016 *	0.299	0.135-0.661	.003 *

Asterisks indicate statistical significance.

* $P < .05$.

analyzed the sequencing data and developed analysis tools. Lili Xia, Hongkang Hu, and Heng Jiang analyzed experimental results. Zhiping Wan, Fei Sheng and Yan Ma assisted in obtaining samples and collected clinical information. Wen Li, Jun Qian and Chun Luo reviewed and edited the manuscript. All authors read and approved the manuscript.

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