ORIGINAL RESEARCH Long non-coding RNA FLJ33360 participates in ovarian cancer progression by sponging miR-30b-3p

This article was published in the following Dove Press journal: OncoTargets and Therapy

Meigin Yang¹ Zhensheng Zhai² Shuang Guo¹ Xiaoxi Li^I Yongxia Zhu¹ Yue Wang¹

¹Department of Gynecology and Obstetrics, People's Hospital of Zhengzhou University, Henan Provincial People's Hospital, Zhengzhou 450000, Henan, People's Republic of China; ²Department of Hepato-Biliary-Pancreatic Surgery, People's Hospital of Zhengzhou University, Henan Provincial People's Hospital, Zhengzho 450000, Henan, People's Republic of China

Correspondence: Yue Wang Department of Gynaeclogy and Obstetrics, People's Hospital of Zhengzhou University, Henan Provincial People's Hospital, No. 7 Weiwu Road, Jinshui District, Zhengzhou 450000, Henan, People's Republic of China Tel +861 352 358 3939 Fax +8 603 716 558 0809 E-mail wangyue0601@163.com



Background: Long noncoding RNAs (lncRNAs) have been reported to play a key role in the development and progression of human malignancies. FLJ33360 is an lncRNA with unknown functions. This study was designed to determine the clinical significance and mechanism of FLJ33360 in ovarian cancer.

Materials and methods: The clinical significance of FLJ33360 in ovarian cancer was determined using the Gene Expression Profiling Interactive Analysis (GEPIA) database, Kaplan-Meier Plotter database, quantitative reverse transcription polymerase chain reaction (qRT-PCR) and statistical analysis. The regulatory relationships between FLJ33360 and miR-30b-3p were explored through bioinformatics, the Gene Expression Omnibus (GEO) database, the ArrayExpress database and meta-analysis. The possible pathways were predicted using Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) analysis. In addition, the key target genes were identified using a protein-protein interaction (PPI) network, the Cancer Genome Atlas (TCGA) database, and correlation analysis.

Results: FLJ33360 expression was significantly downregulated in ovarian cancer tissue (P=0.0011) and was closely associated with International Federation of Gynecology and Obstetrics (FIGO) stage (P=0.027) and recurrence (P=0.002). FLJ33360 may have potential value in detecting ovarian cancer (area under the curve =0.793). Function analysis demonstrated that FLJ33360 can act as a molecular sponge of miR-30b-3p to regulate the expression of target genes that are mainly involved in positive regulation of smooth muscle cell migration, the unsaturated fatty acid metabolic process, and positive regulation of the epithelial to mesenchymal transition. Among these target genes, BCL2 is the hub gene.

Conclusion: FLJ33360 is a potential biomarker for early diagnosis and prognostic assessment in ovarian cancer and may regulate the expression of genes by sponging miR-30b-3p and thus participate in the development of ovarian cancer.

Keywords: ovarian cancer, long noncoding RNAs, FLJ33360, miR-30b-3p, bioinformatics analysis, meta-analysis

Introduction

Ovarian cancer is the deadliest cancer among gynaecological malignancies.¹ This disease is usually asymptomatic and difficult to detect in the early stages.² At present, surgery and chemotherapy are the main treatments for ovarian cancer. However, relapse rates and drug resistance are high, and the prognosis is poor, especially for advanced patients. The 5-year survival rate of ovarian cancer is less than 40%.³ Sensitive early detection indicators and new therapeutic targets are urgently needed to improve the status quo.^{2,4}

OncoTargets and Therapy 2019:12 4469-4480

you hereby accept the Terms. Non-commercial uses of the work are permitted without any further permission from Dove Medical Press Limited, provided the work is properly attributed. For permission for commercial use of this work, please see paragraphs 4.2 and 5 of our Terms (https://www.dovepress.com/terms.php).

Long noncoding RNAs (lncRNAs) are stable noncoding RNAs that are greater than 200 nucleotides in length.⁵ Many lncRNAs are expressed at low levels and have fairly high tissue or cell-type specificity.⁶ MicroRNAs (miRNAs) are a type of endogenous noncoding RNA with approximately 20 nucleotides and various regulatory functions and can inhibit translation and promote degradation of target mRNAs.⁷ Recent research has shown that lncRNAs can regulate gene expression by titrating miRNAs in a phenomenon known as the competing endogenous RNA (ceRNA) hypothesis⁸ (also known as the "sponge" function)⁹ to participate in various biological processes. For example, Grelet et al¹⁰ found that the lncRNA PNUTS can act as a sponge for miR-205 during the epithelial-mesenchymal transition of tumour cells. Chen et al¹¹ found that the lncRNA UICLM acts as a ceRNA for miR-215 to regulate the expression of ZEB2 and thereby control colorectal cancer liver metastasis. However, the specific mechanisms of many lncRNAs remain obscure. FLJ33360 (LINC02145/CTD-2324F15.2) is located on 7p15.2, with the NCBI Gene ID 401172. At present, no reports on related functions have been published.

In this study, we first found that FLJ33360 was significantly downregulated in ovarian cancer tissues. To further explore the clinical significance and potential molecular mechanism of FLJ33360 in ovarian cancer, we conducted indepth studies in combination with qRT-PCR, meta-analysis and bioinformatics. The flow chart is shown in Figure 1.

Materials and methods Human samples

A total of 32 ovarian cancer specimens and 31 normal ovarian specimens were collected from the Department of Obstetrics and Gynecology in People's Hospital of Zhengzhou University between January 2018 and October 2018. Cancer specimens were collected from ovarian cancer patients undergoing cytoreductive surgery, and healthy ovarian tissue was collected from patients undergoing adnexectomy due to myoma or uterine prolapse. The patients ranged in age from 28 to 74 years with a mean age of 55 years. Cancer tissues were collected from primary sites, and the healthy ovarian tissues were collected from the surface epithelium. Fresh tissues were frozen in liquid nitrogen within 5 mins after excision and transferred to a -80 °C freezer. No patients received any pre-operative therapy. All experimental protocols were approved by the Ethical Committee of the People's



Figure I A flow chart of this study.

Hospital of Zhengzhou University, and the clinicians and patients signed the consent forms for the use of their tissues in the study. Histopathology and clinical data were extracted from pathology reports and medical records.

Quantitative reverse-transcription PCR

Total RNA was extracted from the tissues using TRIzol reagent (Invitrogen, Carlsbad, CA, USA). Reverse transcription of 1 µg of total RNA into cDNA was performed using a Thermo Scientific Revert Aid First Strand cDNA Synthesis Kit (Thermo Fisher Scientific, Waltham, MA, USA). Quantitative real-time PCR amplification of cDNA was performed with PowerUpTM SYBRTM Green Master Mix (Applied Biosystems, Foster city, CA, USA). For the Applied Biosystems 7500 System, the thermal cycling conditions were 94 °C for 30 sec, followed by 40 cycles for 94 °C for 5 sec and 60 °C for 30 sec. The expression of FLJ33360 relative to β-actin expression was determined using the 2- $\Delta\Delta$ CT method. The primers used were FLJ33360 (forward) 5'-GCTTCCGATGTGCTGTGGTA -3' and FLJ33360 (reverse) 5'-CTGTCTGGGAGGATG-GATGTC-3', β-actin (forward)

5'-CCTGTACGCCAACACAGTGC-3' and β-actin (reverse) 5'-ATACTCCTGCTTGCTGATCC-3'.

The Cancer Genome Atlas (TCGA), Gene Expression Profiling Interactive Analysis (GEPIA), Kaplan-Meier Plotter databases and data collection

We analysed RNA-Seq data from individuals with ovarian cancer from the Illumina HiSeq RNASeq platform using the TCGA data portal (https://tcga-data.nci.nih.gov/tcga/), which contained 374 ovarian cancer tissues and no healthy ovarian tissue samples. The expression level of the desired RNA was normalized by the Deseq package of the R language for further analysis.

GEPIA (http://gepia.cancerpku.cn/) is an interactive website based on TCGA and Genotype-Tissue Expression (GTEx) project data. We extracted GEPIA data to obtain FLJ33360 expression profiles of various types of human cancer and adjacent normal tissues and validated the differential expression of FLJ33360 in ovarian cancer and healthy ovarian tissues. In addition, GEPIA data were used to analyse differences in FLJ33360 expression at different FIGO stages in ovarian cancer patients. The online database Kaplan-Meier Plotter (KM plotter, http://kmplot.com) was used to assess the association between FLJ33360 expression levels and progression-free survival (PFS), which includes FLJ33360 expression and prognosis information for 614 ovarian cancer patients. The patients were stratified according to the median FLJ33360 expression level (high vs low expression). The hazard ratio (HR) with 95% confidence intervals (CIs) and *P*-value were calculated by a PostgreSQL server. *P*<0.05 was considered significant.

Target gene prediction

Four prediction algorithms, including DIANA IncBase v2¹² (http://diana.imis.athena-innovation.gr/), RegRNA2.0¹³ (http://regrna2.mbc.nctu.edu.tw/detection.html), miRDB¹⁴ (http://mirdb.org/) and IncRNAMap¹⁵ (http://lncrnamap. mbc.nctu.edu.tw/php/), were used to predict miRNAs that interact with FLJ33360. A total of five target prediction algorithms were used to predict the potential target genes of miRNA. The five corresponding prediction algorithms were miRWalk¹⁶ (http://mirwalk.umm.uni-heidelberg.de/), DIANA Tarbase v8¹⁷ (http://carolina.imis.athena-innovation.gr), miRDB (http://mirdb.org/), miRMap¹⁸ (http://mirmap.ezlab.org/) and miRTarBase¹⁹ (http://mirtar base.mbc.nctu.edu.tw/php/search.php). Venn diagrams were used to identify overlapping target genes (http:// bioinformatics.psb.ugent.be/webtools/Venn/).

Selection of microarrays in Gene Expression Omnibus (GEO) and ArrayExpress datasets

Ovarian cancer-related microarrays were downloaded from the National Center of Biotechnology Information (NCBI) GEO (http://www.ncbi.nlm.nih.gov/geo/) and European Bioinformatics Institute's (EBI) ArrayExpress (http:// www.ebi.ac.uk/arrayexpress/) database. The search keywords were as follows: miRNA and mRNA and (malignant* OR cancer OR tumor OR neoplasm* OR carcinoma) and (oophoroma OR ovarian OR OV). The retrieval date was up to October 1, 2018. The screening criteria included (1) corresponding miRNA or mRNA expression in ovarian cancer and normal ovarian samples that can be detected; (2) study samples from human tissue, cell lines, exosomes or body fluids; and (3) more than 3 subjects per group.

Meta-analysis

Stata 14.0 (Stata Corp., College Station, TX, USA) was used for the meta-analysis of miRNA microarrays in this

study. Heterogeneity among the included studies was assessed by the chi-squared and I-squared tests. A fixed effects model was selected when the heterogeneity was low ($I^2 \le 50\%$ and P > 0.05), and when heterogeneity was significant, a random effects model was selected ($I^2 > 50\%$ or $P \le 0.05$). A funnel plot was used to detect a publication bias.

Functional analysis of target genes

We performed Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) analyses using the GluGO plug-in Cytoscape 3.6.1.^{20,21} The STRING database²² (http:// string-db.org) provides a critical assessment and integration of protein-protein interactions (PPIs) by combining a large amount of known and predicted protein-protein association data. We used the STRING website to build the PPI network.

Statistical analysis

Statistical analysis was performed using SPSS 22.0 (IBM, Chicago, IL, USA) and GraphPad Prism 7.0 (GraphPad Software, La Jolla, CA, USA). Student's *t*-test was used to analyse differences between the two groups. The receiver operating characteristic (ROC) curve was used to predict the diagnostic potential of FLJ33360 in ovarian cancer. Pearson correlation analysis was used to study the correlation between FLJ33360 and BCL2 or DLG2. A *P*-value <0.05 was considered significant.

Results

Expression and clinical value of FLJ33360 in ovarian cancer

The GEPIA results showed that FLJ33360 was differentially expressed in a variety of tumours and healthy tissues, with differences in ovarian cancer being particularly pronounced (Figure 2A). A total of 88 healthy ovarian tissues and 426 ovarian cancer tissues were included in the study, and the results showed that FLJ33360 expression in ovarian cancer was significantly downregulated (P<0.05, Figure 2B). In addition, with the progression of the disease FIGO stage, FLJ33360 expression was significantly decreased (P<0.05, Figure 2C). The Kaplan-Meier curve showed that FLJ33360 was closely related to PFS in ovarian cancer patients. Patients with low FLJ33360 expression had a higher recurrence rate than those with high expression (P<0.05, Figure 2D). To further validate this differential expression, we collected 32 ovarian cancer tissues and 31 healthy ovarian tissues. The PCR results showed that FLJ33360 expression was significantly different (P=0.0011, Figure 2E). The correlations between the FLJ33360 expression level and clinical pathological features of ovarian cancer patients are shown in Table 1. The expression level of FLJ33360 has no association with age and tumour size (all P>0.05). Compared with low-grade serous carcinoma and others, FLJ33360 expression was markedly reduced in high-grade serous carcinoma (P=0.027). Compared with the I/II stage, FLJ33360 expression was significantly decreased in the III/IV stage (P=0.027). The ROC curve analysis results showed that the area under the curve (AUC) was 0.793 (95% CI: 0.663–0.923) (P<0.001, Figure 2F).

Prediction of miRNAs that interact with FLJ33360

DIANA IncBase v2, miRDB, RegRNA2.0 and IncRNAMap predicted 37, 152, 83 and 127 miRNAs that may interact with FLJ33360, respectively. A Venn diagram (Figure 3) showed that two intersections of miRNAs are jointly predicted by the four algorithms: miR-30b-3p and miR-4755-3p.

Verification of the differential expression of the two miRNAs in ovarian cancer

To verify the differential expression of miR-30b-3p and miR-4755-3p in ovarian cancer, we performed a metaanalysis of eight eligible microarray datasets in GEO (GSE31568, GSE53829, GSE61741, GSE48485, GSE58517, GSE67329, GSE83693, and GSE76449) and 1 dataset in ArrayExpress (E-MTAB-4667). The characteristics of the selected datasets are shown in Table 2. In total, 222 ovarian cancer samples and 103 control samples were included. In the meta-analysis, the heterogeneity between the miR-30b-3p-related datasets was small $(P=0.325, I^2=12.8\%)$; therefore, the fixed effects model was selected to calculate the combined standard mean difference (SMD). The pooled SMD of miR-30b-3p was 0.44 (95% CI: 0.19–0.69, P=0.001) (Figure 4A). Heterogeneity was observed between the miR-4755-3prelated datasets (P=0.081, $I^2=55.4\%$), and the random effect model was used to calculate the pooled SMD of miR-4755-3p, which was 0.69 (95% CI:-0.25-1.62, P=0.149) (Figure 4B). In addition, the funnel plot showed no significant publication biases in these datasets (Figure 4C and D). The results indicated that miR-30b-3p



Figure 2 The clinical significance of FLJ33360 in ovarian cancer.

Notes: (A) The expression profile of FLJ33360 based on the GEPIA database. (B) Differential expression of FLJ33360 between ovarian cancer and healthy ovarian tissues based on the GEPIA database. *P<0.05. (C) Differential expression of FLJ33360 in stage II, stage III and stage IV ovarian cancer. (D) The relationship between FLJ33360 expression and PFS of ovarian cancer patients based on the KM Plotter. (E) Differential expression of FLJ33360 based on qRT-PCR. (F) ROC curve of FLJ33360 based on qRT-PCR.

Abbreviations: OV, ovarian cancer; GEPIA, gene expression profiling interactive analysis; PFS, progression-free survival; KM, Kaplan-Meier; qRT-PCR, quantitative reverse transcription polymerase chain reactions; ROC, receiver operating characteristic.

was significantly upregulated in ovarian cancer, while the difference in miR-4755-3p expression was not significant. We selected miR-30b-3p as an important miRNA that competitively binds to FLJ33360 in ovarian cancer.

Acquisition of target genes in ovarian cancer MiRDB, miRMap, miRWalk, DIANA Tarbase v8 and miRTarBase predicted 1096, 6755, 4570, 394 and 667 mRNAs targeted by miR-30b-3p, respectively. The Venn

Clinicopathological features	Cases	FLJ33360 expression	Р
Age (years)			
≤50	9	0.66±0.58	0.442
>50	23	0.50±0.52	
Histological type			
Low-grade serous carci-	7	0.93±0.48	0.027
noma and others			
High-grade serous	25	0.44±0.50	
carcinoma			
Clinical stage			
1/11	5	1.02±0.58	0.027
III/IV	27	0.46±0.48	
Tumour size (cm)			
<10	17	0.58±0.58	0.725
≥10	15	0.51±0.49	

Table	I	The	correlations	between	FLJ33360	expression	and
clinicop	bat	holog	ical paramet	ers of ova	rian cancer	[•] patients	



Figure 3 A flow chart to screen the miRNAs based on Venn diagrams.

diagram (Figure 5) showed that at least two algorithms repeatedly predicted 3355 target genes. To further identify target genes that are differentially expressed in ovarian

Table 2 Characteristics of the datasets included in the meta-analysis

Series Country Year Platform Sample source **OV** samples Normal samples GSE31568 Germany 2011 GPL9040 Peripheral blood 15 19 GSE53829 China 2014 GPL18138 Tissue 39 14 GSE61741 Germany 2014 GPL9040 Peripheral blood 24 26 GSE48485 2014 GPL14943 Peripheral blood 5 5 Germany 2015 GPL18402 Urine 5 5 GSE58517 China USA 2015 GPL8179 Cell 5 2 GSE67329 12 GSE76449 USA 2016 GPL19117 Cell 2 GSE76449 USA 2016 GPL19117 12 2 Exosomes GSE83693 South Korea 2017 GPL22079 Tissue 4 4 101 24 E-MTAB-4667 Italy 2017 NR Peripheral blood

Abbreviations: OV, ovarian cancer; NR, not reported.

cancer, we extracted a total of 6434 mRNAs that are differentially expressed in ovarian cancer and healthy ovarian tissues from the dataset GSE14407. The expression levels of 696 mRNAs showed greater than 2-fold decreases (adj. P-value<0.05, log2FC<-2.0). The above 3355 mRNAs and 696 mRNAs were combined to obtain 104 potential genes that were targeted by miR-30b-3p in ovarian cancer.

Functional analysis of target genes in ovarian cancer

GO analysis provides a biological functional interpretation of a large number of genes derived from genomic studies. GO analysis of the 104 target genes revealed some of the biological processes involved, such as positive regulation of smooth muscle cell migration, the unsaturated fatty acid metabolic process, and positive regulation of the epithelial to mesenchymal transition (Table 3, Figure 6A). The enriched KEGG pathways included the AGE-RAGE signalling pathway in diabetic complications and colorectal cancer (Table 4, Figure 6B). The PPI network revealed interactions between target genes (Figure 7A). Among them, BCL2 and DLG2 are hub genes. Based on the GEPIA database, we examined the expression levels of these two genes. BCL2 and DLG2 expression levels were both significantly downregulated in ovarian cancer (P<0.05, Figure 7B and C). In addition, we extracted the expression levels of FLJ33360, BCL2 and DLG2 in ovarian cancer patients in the TCGA database. Pearson correlation analysis results showed a positive correlation between FLJ33360 and BCL2 (r=0.134, P=0.014, Figure 7D) but no significant correlation between FLJ33360 and DLG2 (r=-0.032, P=0.555, Figure 7E).



Figure 4 Meta-analysis of the differential expression of miR-30b-3p and miR-4755-3p between the ovarian cancer group and the control group based on GEO and ArrayExpress datasets. Notes: (A) Forest plot of studies evaluating the SMD of miR-30b-3p (a fixed effects model). (B) Forest plot of studies evaluating the SMD of miR-4755-3p (a random effects model). (C) Funnel plot of miR-30b-3p-related datasets. (D) Funnel plot of miR-4755-3p-related datasets. (D) Funnel plot of miR-4755-3p (a random effects model). (C) Funnel plot of miR-30b-3p-related datasets. (D) Abbreviations: GEO, gene expression onnibus, SMD, standard mean difference.



Figure 5 A flow chart to screen the mRNAs based on Venn diagrams.

Discussion

The high recurrence rate and mortality of ovarian cancer are major problems worldwide. New biomarkers and therapeutic targets are urgently needed to improve the status quo.23 The mechanism and function of lncRNA are increasingly being explored and elucidated. Many studies have shown that lncRNAs containing sequences similar to their target miRNAs can be used to isolate miRNAs from mRNAs to further regulate gene expression and participate in the development and progression of various types of tumours.²⁴ The discovery of aberrantly expressed lncRNAs and related signalling pathways is important for improving the status of ovarian cancer.²⁵ In this study, we first determined the change in expression of lncRNA FLJ33360. A large amount of data indicates that FLJ33360 expression is significantly downregulated in ovarian cancer, especially in high-grade serous carcinoma. As the disease stage progresses, the downregulation of FLJ33360 is significant. In addition, low FLJ33360 expression is associated with high recurrence in ovarian cancer patients and may serve as a potential biomarker for

Table 3	Enrichment	GO	terms	of the	target	genes
---------	------------	----	-------	--------	--------	-------

prognostic evaluation of ovarian cancer. FLJ33360 may also have potential for detecting ovarian cancer. In our follow-up study, we will investigate whether FLJ33360 is differentially expressed in blood between ovarian cancer patients and normal controls.

To explore the underlying pathways of FLJ33360 in ovarian cancer to further understand its mechanism, based on the interaction between miRNAs and lncRNAs, this study screened two miRNAs that interact with the IncRNA FLJ33360: miR-30b-3p and miR-4755-3p. Our study has shown that FLJ33360 is significantly downregulated in ovarian cancer. Based on the meta-analysis results, we excluded miR-4755-3p from the angle of ceRNA role of lncRNA. miR-30b-3p has been reported to play an important role in human cancer.²⁶ This study found for the first time that miR-30b-3p is significantly upregulated in ovarian cancer, and that FLJ33360 downregulates the expression of target genes by competitive binding of miR-30b-3p. These target genes are enriched in many different biological processes and signalling pathways, including positive regulation of smooth muscle cell migration, the unsaturated fatty acid metabolic process, and positive regulation of the epithelial to mesenchymal transition. Notably, FLJ33360 was not identified among the validated targets of miR-30b-3p, possibly because FLJ33360 belongs to the IncRNA category and is not present in the mRNA library of the miRNA-predicted mRNA module in the prediction algorithms used in this study. BCL2 is the hub gene of these target genes. The expression of BCL2 is upregulated in most tumour tissues and promotes tumour development by inhibiting apoptosis.²⁷ However, many studies have shown a decrease in BCL2 expression in ovarian cancer.^{28,29,30,31} which is consistent with the results of this study. The mechanism of action of BCL2 in ovarian cancer is

GO ID	Term	Ontology	Count	P-value
GO:0014911	positive regulation of smooth muscle cell migration	ВР	4	4.66E-05
GO:0006582	melanin metabolic process	BP	3	1.91E-04
GO:0043244	regulation of protein complex disassembly	BP	5	2.42E-04
GO:0033559	unsaturated fatty acid metabolic process	BP	5	3.57E-04
GO:0048821	erythrocyte development	BP	3	5.62E-04
GO:0006692	prostanoid metabolic process	BP	3	1.41E-03
GO:0010718	positive regulation of epithelial to mesenchymal transition	BP	3	I.74E-03
GO:0050918	positive chemotaxis	BP	3	7.01E-03
GO:0002088	lens development in camera-type eye	BP	3	7.56E-03
GO:0048332	mesoderm morphogenesis	BP	3	7.85E-03

Abbreviations: GO, gene ontology; BP, biological process.



Figure 6 GO and KEGG analysis of potential target genes using the Cytoscape plug-in GluGO. Notes: (A) A functional network of GO terms. (B) KEGG pathways displayed as an interaction network. Abbreviations: GO, gene ontology; KEGG, Kyoto encyclopedia of genes and genomes.

Table 4	4 KEGG	pathway	enrichment	analysis	of the	target	genes
---------	--------	---------	------------	----------	--------	--------	-------

KEGG ID	KEGG term	Count	P-value	Gene symbol
KEGG:04933	AGE-RAGE signaling pathway in diabetic complications	4	6.45E-03	AKT3, BCL2, TGFBR2, THBD
KEGG:05210	Colorectal cancer	3	1.68E-02	AKT3, BCL2, TGFBR2

Abbreviation: KEGG, Kyoto Encyclopedia of genes and genomes.

complex.³² On the one hand, regulation of the BCL2 family, especially individual proteins, in ovarian cancer cell apoptosis appears to be tissue-specific.^{31,33} On the other hand, in addition to acting as a regulator of apoptosis in ovarian

cancer, BCL2 may have other effects,²⁹ such as inhibiting angiogenesis^{28,30} and preventing cells from entering the S phase, leading to specific growth inhibition.³⁰ Other researchers have also pointed out that frequent apoptosis



Figure 7 BCL2 is the hub gene of potential target genes.

Notes: (A) The PPI network of the target genes. Nodes represent gene-encoded proteins. Connections between nodes represent the relationship between proteins. A bold line indicates a higher confidence level. (B) Differential expression of BCL2 between ovarian cancer and healthy ovarian tissues based on the GEPIA database. *P<0.05. (C) Differential expression of DLG2. *P<0.05. (D) The correlation between FLJ33360 and BCL2 based on the TCGA database. (E) The correlation between FLJ33360 and DLG2.

Abbreviations: PPI, protein-protein interaction; GEPIA, gene expression profiling interactive analysis; TCGA, the cancer genome atlas.

in ovarian cancer is a sign of high conversion of invasive malignant cells and may play a role in eliminating cells that carry DNA damage and in transforming these cells into cancer cells.³² Taken together, FLJ33360 may regulate the expression of target genes such as BCL2 to participate in different biological processes via sponging miR-30b-3p in ovarian cancer.

Conclusion

We found that FLJ33360 expression was significantly downregulated in ovarian cancer and was closely related to the clinical stage and recurrence of the disease. FLJ33360 appears to be a potential biomarker for early diagnosis and prognostic assessment in ovarian cancer. FLJ33360 may be involved in multiple biological pathways in ovarian cancer by competitively binding to miR-30b-3p, thus warranting in-depth studies with in vitro and in vivo experiments. This study provides new insights into the clinical significance and mechanism of action of FLJ33360 in ovarian cancer and may reveal new targets for future research and treatment.

Abbreviations list

OV, ovarian cancer; lncRNAs, long noncoding RNAs; qRT-PCR, quantitative reverse transcription polymerase chain reactions; GEPIA, Gene Expression Profiling Interactive Analysis; ROC, receiver operating characteristic; GEO, Gene Expression Omnibus; ceRNA, competing endogenous RNA; GO, gene ontology; KEGG, kyoto encyclopedia of genes and genomes; PPI, protein-protein interaction; TCGA, the cancer genome atlas; GTEx, genotype-tissue expression; KM, Kaplan-Meier; PFS, progression-free survival; HR, hazard ratio; CI, confidence interval; SMD, standard mean difference; BP, biological process.

Acknowledgment

This study was supported by research funding from the Science and Technology Commission of Henan Province (Grant number: 162102310174).

Disclosure

The authors report no conflicts of interest in this work.

References

- Yokoi A, Matsuzaki J, Yamamoto Y, et al. Integrated extracellular microRNA profiling for ovarian cancer screening. *Nat Commun.* 2018;9(1):4319. doi:10.1038/s41467-018-06434-4
- Pisanic TR 2nd, Cope LM, Lin SF, et al. Methylomic analysis of ovarian cancers identifies tumor-specific alterations readily detectable in early precursor lesions. *Clin Cancer Res.* 2018;24(24):6536–6547. doi:10.1158/1078-0432.CCR-18-1199
- Heindl A, Khan AM, Rodrigues DN, et al. Microenvironmental niche divergence shapes BRCA1-dysregulated ovarian cancer morphological plasticity. *Nat Commun.* 2018;9(1):3917. doi:10.1038/s41467-018-06130-3
- 4. Blyuss O, Burnell M, Ryan A, et al. Comparison of longitudinal CA125 algorithms as a first-line screen for ovarian cancer in the general population. *Clin Cancer Res.* 2018;24(19):4726–4733. doi:10.1158/1078-0432.CCR-18-0208
- Leisegang MS. LET's sponge: how the lncRNA PFL promotes cardiac fibrosis. *Theranostics*. 2018;8(4):874–877. doi:10.7150/ thno.23364
- Krause HM. New and prospective roles for lncRNAs in organelle formation and function. *Trends Genet*. 2018;34(10):736–745. doi:10.1016/j.tig.2018.06.005
- Rzepiela AJ, Ghosh S, Breda J, et al. Single-cell mRNA profiling reveals the hierarchical response of miRNA targets to miRNA induction. *Mol Syst Biol.* 2018;14(8):e8266. doi:10.15252/ msb.20188266
- Wen Q, Zhao L, Wang T, et al. LncRNA SNHG16 drives proliferation and invasion of papillary thyroid cancer through modulation of miR-497. *Onco Targets Ther.* 2019;12:699–708. doi:10.2147/OTT. S186923
- Kim J, Abdelmohsen K, Yang X, et al. LncRNA OIP5-AS1/cyrano sponges RNA-binding protein HuR. *Nucleic Acids Res.* 2016;44 (5):2378–2392. doi:10.1093/nar/gkw017
- Grelet S, Link LA, Howley B, et al. A regulated PNUTS mRNA to IncRNA splice switch mediates EMT and tumour progression. *Nat Cell Biol.* 2017;19(9):1105–1115. doi:10.1038/ncb3450
- Chen DL, Lu YX, Zhang JX, et al. Long non-coding RNA UICLM promotes colorectal cancer liver metastasis by acting as a ceRNA for microRNA-215 to regulate ZEB2 expression. *Theranostics*. 2017;7 (19):4836–4849. doi:10.7150/thno.20942
- Paraskevopoulou MD, Vlachos IS, Karagkouni D, et al. DIANA-LncBase v2: indexing microRNA targets on non-coding transcripts. *Nucleic Acids Res.* 2016;44(D1):D231–238. doi:10.1093/nar/gkv1270

- Huang HY, Chien CH, Jen KH, Huang HD. RegRNA: an integrated web server for identifying regulatory RNA motifs and elements. *Nucleic Acids Res.* 2006;34:W429–W434. doi:10.1093/nar/gkl333
- Wong N, Wang X. miRDB: an online resource for microRNA target prediction and functional annotations. *Nucleic Acids Res.* 2015;43 (D1):D146–D152. doi:10.1093/nar/gku1104
- Chan WL, Huang HD, Chang JG. lncRNAMap: a map of putative regulatory functions in the long non-coding transcriptome. *Comput Biol Chem.* 2014;50:41–49. doi:10.1016/ j.compbiolchem.2014.01.003
- Sticht C, De La Torre C, Parveen A, Gretz N. miRWalk: an online resource for prediction of microRNA binding sites. *PLoS One*. 2018;13(10):e0206239. doi:10.1371/journal.pone.0206239
- Karagkouni D, Paraskevopoulou MD, Chatzopoulos S, et al. DIANA-TarBase v8: a decade-long collection of experimentally supported miRNA-gene interactions. *Nucleic Acids Res.* 2018;46(D1):D239– D245. doi:10.1093/nar/gkx1141
- Vejnar CE, Blum M, Zdobnov EM. miRmap web: comprehensive microRNA target prediction online. *Nucleic Acids Res.* 2013;41: W165–168. doi:10.1093/nar/gkt430
- Chou CH, Shrestha S, Yang CD, et al. miRTarBase update 2018: a resource for experimentally validated microRNA-target interactions. *Nucleic Acids Res.* 2018;46(D1):D296–D302. doi:10.1093/nar/gkx1067
- Bindea G, Mlecnik B, Hackl H, et al. ClueGO: a Cytoscape plug-in to decipher functionally grouped gene ontology and pathway annotation networks. *Bioinformatics*. 2009;25(8):1091–1093. doi:10.1093/bioinformatics/btp101
- 21. Saito R, Smoot ME, Ono K, et al. A travel guide to Cytoscape plugins. *Nat Methods*. 2012;9(11):1069–1076. doi:10.1038/ nmeth.2212
- 22. Szklarczyk D, Morris JH, Cook H, et al. The STRING database in 2017: quality-controlled protein–protein association networks, made broadly accessible. *Nucleic Acids Res.* 2017;45(D1):D362–D368. doi:10.1093/nar/gkw937
- van Den Brand D, Mertens V, Massuger LFAG, Brock R, Society BRJJocrojotCR. siRNA in ovarian cancer - Delivery strategies and targets for therapy. *J Control Release*. 2018;283::45–58.
- 24. genetics LEJFi. Sequence determinants for nuclear retention and cytoplasmic export of mRNAs and lncRNAs. *Front Genet*. 2018;9::440.
- 25. Teschendorff AE, Lee SH, Jones A, et al. HOTAIR and its surrogate DNA methylation signature indicate carboplatin resistance in ovarian cancer. *Genome Med.* 2015;7:108. doi:10.1186/s13073-015-0233-4
- 26. Wang Q, Hui H, Guo Z, et al. ADAR1 regulates ARHGAP26 gene expression through RNA editing by disrupting miR-30b-3p and miR-573 binding. *RNA*. 2013;19(11):1525–1536. doi:10.1261/ rna.041533.113
- Klein S, Abraham M, Bulvik B, et al. CXCR4 promotes neuroblastoma growth and therapeutic resistance through miR-15a/16-1-Mediated ERK and BCL2/Cyclin D1 pathways. *Cancer Res.* 2018;78(6):1471–1483. doi:10.1158/0008-5472.CAN-17-0454
- Schuyer M, Van-Der-Burg M, Henzen-Logmans S, et al. Reduced expression of BAX is associated with poor prognosis in patients with epithelial ovarian cancer: a multifactorial analysis of TP53, p21, BAX and BCL-2. *Br J Cancer*. 2001;85(9):1359–1367. doi:10.1054/bjoc.2001.2101
- Chan WY, Cheung KK, Schorge JO, et al. Bcl-2 and p53 protein expression, apoptosis, and p53 mutation in human epithelial ovarian cancers. *Am J Pathol.* 2000;156(2):409–417. doi:10.1016/S0002-9440(10)64698-6
- Baekelandt M, Kristensen GB, Nesland JM, Tropé CG, Holm R. Clinical significance of apoptosis-related factors p53, Mdm2, and Bcl-2 in advanced ovarian cancer. *J Clin Oncol.* 1999;17(7):2061. doi:10.1200/JCO.1999.17.7.2061

- Marone M, Scambia G, Mozzetti S, et al. bcl-2, bax, bcl-XL, and bcl-XS expression in normal and neoplastic ovarian tissues. *Clin Cancer Res.* 1998;4(2):517–524.
- 32. de la Torre FJ, Garcia A, Gil-Moreno A, et al. Apoptosis in epithelial ovarian tumours Prognostic significance of clinical and histopathologic factors and its association with the immunohistochemical expression of apoptotic regulatory proteins (p53, bcl-2 and bax). *Eur J Obstet Gynecol Reprod Biol*. 2007;130(1):121–128. doi:10.1016/j.ejogrb.2005.11.048
- 33. Diebold J, Baretton G, Felchner M, et al. bcl-2 expression, p53 accumulation, and apoptosis in ovarian carcinomas. Am J Clin Pathol. 1996;105(3):341–349. doi:10.1093/ajcp/105.3.341

OncoTargets and Therapy

Publish your work in this journal

OncoTargets and Therapy is an international, peer-reviewed, open access journal focusing on the pathological basis of all cancers, potential targets for therapy and treatment protocols employed to improve the management of cancer patients. The journal also focuses on the impact of management programs and new therapeutic

Submit your manuscript here: https://www.dovepress.com/oncotargets-and-therapy-journal

Dovepress

agents and protocols on patient perspectives such as quality of life, adherence and satisfaction. The manuscript management system is completely online and includes a very quick and fair peer-review system, which is all easy to use. Visit http://www.dovepress.com/ testimonials.php to read real quotes from published authors.