Comparative Transcriptome Provides a Systematic Perspective on Epstein–Barr Virus-Associated Gastric Carcinoma Cell Lines

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Purpose: Epstein–Barr virus (EBV) is widely recognised to cause various tumours, and EBV-associated gastric carcinoma (EBVaGC) is a special type of GC. It has obviously different clinical features and pathological manifestations from EBV-negative gastric carcinoma, but its progression remains elusive. The underlying cancer progression of viral infection detected by genome-wide transcriptome analysis has been demonstrated in numerous diseases.

Methods: We performed comparative RNA sequencing to identify gene expression signatures between GC and EBVaGC cell lines. The differentially expressed (DE) genes were analysed using gene ontology and pathway enrichment.

Results: A total of 4438 DE mRNAs, 3650 DE long non-coding RNAs (lncRNAs), and 248 DE circular RNAs (circRNAs) were detected in GC cells after EBV infection, most of which were highly related to oncogenesis. Likewise, EBV-coding RNA and non-coding RNA were also well-supplemented in EBVaGC. According to bioinformatics, DE mRNAs may contribute to the completion of EBV-infected host cells and modulate mitosis. Binding to actin and participating in adherens junctions to promote contact between the virus and cells are a potential function of DE lncRNAs. The roles of DE circRNAs were enriched in DNA repair and protein modification, and a typical example of this is acting as an miRNA sponge. The establishment of a circRNA-miRNA-mRNA network helps to determine the key elements in the progression of EBVaGC.

Conclusion: This study is the first to systematically reveal the transcriptome landscape of EBVaGC, which will provide an essential resource for genomic, genetic, and molecular mechanisms in the future.

Keywords: transcriptome, RNA-sequencing, bioinformatics, regulatory network, EBV-associated gastric cancer

Introduction

Epstein–Barr virus (EBV), formally known as human gammaherpesvirus 4, is one of the most common viruses in humans, infecting nearly 90% of adults and providing adaptive immunity. EBV infection can lead to the occurrence of a variety of diseases, such as B-cell-related diseases including infectious mononucleosis and Burkitt lymphoma, and epithelial-related diseases such as nasopharyngeal carcinoma and gastric carcinoma (GC).¹ EBV encoded small RNA (EBER) was detected by in situ hybridisation (ISH) in gastric cancer tissue, which was defined as EBV-associated gastric carcinoma (EBVaGC). A number of clinical
studies have shown that EBVaGC accounts for 10% of global GC cases, with nearly 95,000 new cases of EBVaGC occurring worldwide each year. In 2014, the Cancer Genome Atlas (TCGA) working group reported a new classification of GC divided into four subtypes: Epstein–Barr virus positive, microsatellite unstable (MSI), genomically stable (GS), and chromosomal instability (CIN), further emphasising that EBVaGC is an independent molecular type of GC. EBVaGC compared with EBV-negative gastric carcinoma (EBVnGC) has significantly different clinicopathological features: mostly occurring in males and younger individuals, with better prognosis, located in the cardia and gastric fundus, showing lymphocyte infiltration, and having a general diffuse histological type. The aforementioned cases indicate that EBVaGC may have a unique cancer progression in EBVnGC. However, there is no specific treatment for EBVaGC, and the molecular mechanisms remain largely unknown.

Noncoding RNA (ncRNA) is a class molecule that cannot be translated into protein, because it has no function and for a long time was known as “junk RNA”—a product of spurious transcription. There is increasing evidence that ncRNAs are involved in almost every aspect of biological and pathological processes, and many ncRNA have been redefined as functional, including microRNAs (miRNAs), long non-coding RNAs (lncRNAs), and circular RNAs (circRNAs). In our previous miRNA sequencing studies, it was found that EBV miRNAs could promote the metastasis of EBVaGC by targeting APC and DKK1 and activating the Wnt signaling pathway. Moreover, differential expression of EBV miRNA was significantly associated with lymph node metastasis and 5-year overall survival. To the best of our knowledge, the expression profile of the whole transcriptome has not yet been reported in EBVaGC, particularly in lncRNAs and circRNAs, and their relationship to mRNA. Therefore, we first infected human GC cell line AGS with EBV to construct the AGS-EBV cell line and then sequenced the expression profiles of miRNAs, lncRNAs, and circRNAs in the AGS and AGS-EBV groups. The functions of the RNA were analysed using Gene Ontology (GO), Kyoto Encyclopaedia of Genes and Genomes (KEGG), and Reactome. Finally, the circRNA-miRNA-mRNA network was constructed to identify key genes that might contribute to the progression of EBVaGC.

Materials and Methods
Cell Culture and RNA Isolation
AGS is an EBV-negative human GC cell line. The AGS-EBV cell line was established from AGS cells co-cultivated with Akata cells using the cell-to-cell contact method described as AGS cell infection with a recombinant EBV strain. All cells were cultured in RPMI-1640 medium (Gibco, USA) containing 10% foetal bovine serum (FBS; Gibco) at 37 °C with 5% CO₂. Total RNA was extracted from cultured cells using TRIzol reagent (Invitrogen, Carlsbad, CA, USA) according to the manufacturer’s instructions. The RNA extracts were stored at -80 °C.

Successful Establishment of AGS-EBV Model
We cultured Akata cells that could be modified to produce recombinant EBV, and the recombinant viral DNA contained the neomycin resistance gene (Neo) and green fluorescent protein (GFP) for easy screening and observation. We then used goat anti-IgG to induce EBV in Akata cells from the latent to active state for 24 h, followed by co-incubation with AGS cells for 48 h. Finally, Akata cells were washed with RPMI, and 500 μg/mL G418 (Gibco) was added to RPMI to select for retention of the EBV episome. GFP in cells was observed by fluorescence microscopy to identify the EBV-positive AGS cell line (AGS-EBV).

RNA Sequencing
The reads were first mapped to the latest UCSC transcript set using Bowtie2 version 2.1.0, and the gene expression level was estimated using RSEM v1.2.15. Whole-genome sequencing reads were mapped to the EBV reference genome (NC_007605). For lncRNA expression analysis, we used the transcripts set from Lncipedia (http://www.lncipedia.org). TMM (trimmed mean of M-values) was used to normalise gene expression. Differentially expressed (DE) genes were identified using edgeR. Genes showing altered expression with p < 0.05 and more than 1.5-fold changes were considered differentially expressed. For circRNA expression analysis, the reads were mapped to the genome using STAR, and DCC was used to identify the cirRNA and to estimate circRNA expression. TMM (trimmed mean of M-values) was used to normalise gene expression. DE genes were identified using EdgeR.
GO Analysis

GO analysis (http://www.geneontology.org) was used to set up the gene annotations. The DE transcriptome was classified into GO terms, including biological process (BP), cellular component (CC), and molecular function (MF). Fisher’s exact test was applied for the GO analysis with a significant P-value calculated, and FDR was used to correct the P-values.

KEGG and Reactome Pathway Analysis

Kyoto Encyclopaedia of Genes and Genomes (KEGG) (https://www.kegg.jp) and Reactome (http://reactome.org) pathway analyses were utilized to describe the genes’ attributes. Fisher’s exact test was used to select the significant pathway, and the threshold of significance was defined as FDR < 0.05. To assess the main functional pathways of EBVaGC, DE genes were assessed by GO annotation, KEGG, and Reactome pathway analyses using the cluster Profiler package in R.⁹

GO Tree

The GO is structured as a directed acyclic graph, and each term defines the relationships to one or more other terms. The GO tree was built using the GO gene annotation terms. In the GO analysis based on up-down DE genes, significant GO items (p < 0.01) were selected to construct a topology plot. This was a more convenient visual method to summarise the influence functions in this study.¹⁰

Real-Time Quantitative Polymerase Chain Reaction Verification of DE Genes

Based on the comprehensive analysis of sequencing data and existing databases, qRT-PCR was performed to verify the transcriptome sequencing results of 12 genes that may be involved in viral infection and regulation of cell growth, cell cycle, and cell apoptosis. Premier software (version 6.0) was used to design the qPCR primers, which were synthesised by Sangon Biotech Company. β-actin was used as an internal reference gene, as shown in the Supplementary Material.

Prediction of circRNA-miRNA and miRNA-mRNA Interaction

MiRanda (http://www.microrna.org) was used to predict the miRNA target of the circRNA, and the DE circRNAs within all comparisons were annotated in detail with the circRNA-miRNA interaction information. miRNA target genes were obtained from miRBase (www.mirbase.org) to construct the miRNA-mRNA interactions. R software (http://www.r-project.org) was used to generate the network.

Statistical Analysis

All statistical analyses were performed using SPSS software (version 20.0; IBM, SPSS, USA). One-way analysis of variance (ANOVA) was used to analyse the difference in expression levels between groups, followed by Student’s t-test. Linear correlation analysis was used to verify the consistency of trends between the q-PCR and RNA-seq results. A P value ≤ 0.05 and fold change of 1.5 were considered statistically significant.

Results

Whole Transcriptome Sequence Analysis

First, whole-genome sequencing reads were mapped to the human reference genome (UCSC hg19) and the EBV reference genome (NC_007605). The workflow of the overall data analysis is shown in Figure s1. Total RNA was extracted from AGS and AGS-EBV, and valid reads were obtained after rRNA depletion. Reads were mapped to the genome using TopHat2¹¹ to analyse the expression of mRNAs and lncRNAs. The remaining unmapped reads were subjected to circRNA prediction algorithm to identify the circRNA expression profile, and then the circRNA-miRNA network was constructed by the pair’s prediction between circRNA and its potential target miRNA. Finally, all DE mRNA, lncRNA, and circRNA functional enrichment analyses were performed by GO, and KEGG and Reactome Function enrichment analysis was performed to predict potential biological functions of the differentially expressed genes.

In this experimental process, we found that most reads were mapped to the coding region, untranslated region (UTR), and intronic regions, while reads mapped to intergenic regions had the lowest abundance (Figure 1A). Interestingly, perhaps due to EBV infection stress, the total mapped read of AGS-EBV cells was higher (9.9 x 10⁷ vs 8.0×10⁷) than that in the AGS cells, and also appeared in separate region reads. Although the number of reads varied, the dataset distribution had a great degree of standardisation. As shown in Figure 1B, the median of log2 ratios among the six samples was almost the same after normalisation, indicating that the expression distribution was closed. Furthermore, based on their expression value (fragments per kilobase of transcript per million or FPKM) by mapping all transcripts to the reference
genome, we created a circos-plot to present the DE mRNAs, lncRNAs, and circRNAs between AGS-EBV and uninfected cells (Figure 1C).

**Differential Expression of mRNA in Response to EBV Infection**

The mRNA expression profile showed distinct differences between EBV-infected and uninfected cells in cluster analysis, and the volcano plot clearly displayed the variations (Figure 2A and B). A total of 4438 DE mRNAs were detected in AGS-EBV; 2389 mRNAs were up-regulated, 2049 were down-regulated, and the top 20 DE mRNAs are listed in detail (Supplementary Data).

Bioinformatics analysis has been widely used in tumour research as a crucial tool for exploring gene function and characteristics. Thus, GO, KEGG, and Reactome pathway enrichment analyses were performed to explore the roles of DE mRNAs in AGS-EBV. The GO terms showed the most remarkable biological processes in DE mRNAs, including mitotic nuclear division, negative regulation of the cell cycle, and chromosome segregation. Similarly, 103 DE mRNAs with the highest number of cellular components were mainly found in the nuclear chromosome, and 79 with chromatin binding function were also enriched in the molecular functional annotation region (Figure 2C). Among them, H3 histone family 3A (H3F3A) is one of the 79 genes that we hypothesised to have chromatin binding function and has been confirmed to modify chromatin status and promote tumour cell migration through intronic regulation. Further investigations of cellular components showed that the
nucleoplasm and nucleolus were the core processes of the GO tree (Figures 2D, s2A and B). KEGG and Reactome pathway analysis suggested that the HTLV-I infection pathway, viral carcinogenesis regions, transcriptional misregulation in cancer, cell cycle, cell cycle (mitotic), M phase, and cellular responses to stress were most enriched among the DE
mRNAs (Figure 2E–H). Based on the above results, it is reasonable to speculate that after EBV infection with AGS, DE mRNA is mainly involved in nuclear and chromosome modification, and the integration of the host genome further interferes with the division of EBVaGC cells.

Putative Involvement of the Viral mRNA in Tumour Carcinogenesis

By focusing on the EBVaGC cell line, we not only detected the host cell’s mRNA but also successfully obtained 91 transcripts from the EBV genome, all of which were highly expressed, with the highest mRNA expression being BXL2F (9128 reads per kilobase per million (RPKM)) (Figure s3). BXL2F gene description is envelope glycoprotein H, which is a conserved herpesvirus gene product functionally implicated in the penetration of the virus into the host cell.13 Compared with those previously reported in an EBVaGC tissues study,14 there were 15 additional viral transcripts (RPMS1, LF1, LF2, LF3, orlP, orlLyt, LMP-2B, Wp, Cp, BWRF1, TR, EBNA-LP, Qp, EBER1, EBER2) detected in AGS–EBV cells, with a medium expression level of 396.64 RPKM in all mRNAs. Four of the most expressed genes, RPMSI (1952.3 RPKM), LF2 (1388.6 RPKM), LF1 (776.0 RPKM), and LF3 (670 RPKM), were more than twice as high as those of latent membrane protein 2A (LMP2A, 284.7 RPKM) and Epstein–Barr nuclear antigen −1 (EBNA-1, 270.0 RPKM).

Distinct Expression Profile of lncRNA in AGS-EBV

Over 70% of mammalian genomes are transcribed into non-coding RNAs (ncRNAs), while only 1–2% of mammalian genomes are transcribed into protein-encoded RNAs.15 According to the latest version of the GenCode database, the human genome contains 16,000 lncRNAs; combining data from other species, there are more than 40,000 known lncRNAs. In the present study, 15,590 lncRNAs were identified in AGS-EBV and 3650 DE lncRNAs were screened, of which 1954 were up-regulated and 1696 were down-regulated (Figure 3A and B, Supplementary Data). Since the length of lncRNAs is usually longer than 200nt, we compared the assembled transcriptome with the known human gene annotation and found that DE lncRNAs have multiple isoforms. Furthermore, when EBV entered AGS cells, more lncRNA isoforms were produced, with 2213 additional lncRNAs with an abundance of 449 and 9820 RPKM (Figure 3C).

Expression Profiles and Integrated Screening of circRNAs in EBV-Infected and Uninfected AGS

Unlike the well-known linear RNAs, circRNAs do not have a 5’ cap and 3’ polyadenylated tail and form covalently closed single-stranded RNAs. Since there are few studies on circRNA in EBVaGC, we conducted a more comprehensive analysis. A total of 8426 circRNAs were identified in AGS-EBV, with lengths ranging from 400nt to 500nt, and the longest being 4985nt (Figure 4A). Most circRNAs consisted of protein-coding exons, whereas smaller fractions aligned with intronic, intergenic, antisense, lincRNA, 5’UTR, and 3’UTR (Figure 4B). The abundance of circRNA was lower than that of mRNA and lncRNA; only a quarter of circRNA had more than two back-spliced reads, either AGS-EBV or AGS (Figure 4C and D). Analysis of the number of circRNAs from their host genes revealed that one gene could produce multiple circRNAs, which was consistent with a previous report (Figure 4E).16 A striking example is the oncogene baculoviral IAP repeat containing 6 (BIRC6), which may generate 24 distinct circRNAs (at least two unique back-spliced reads).

A total of 248 DE circRNAs were detected in AGS after EBV infection, with 124 up-regulated and down-regulated circRNAs (Figure 5A and B, Supplementary Data). GO and Pathway analysis of circRNA showed the top 10 enriched terms that may be associated with the mechanism of AGS-EBV (Figure 5C–G). The most meaningful GO terms according to KEGG and Reactome pathway analyses were related to
DNA repair and cell mitosis, such as “DNA repair”, “ligase activity”, “chromatin binding”, “nuclear envelope”, “nuclear body”, and “ubiquitin mediated proteolysis pathway”. In the circRNA GO tree, “protein modification by small protein conjugation or removal” and “protein modification by small protein conjugation” were shown to be the core molecular

Figure 3 Distinct expression profile of IncRNA in AGS-EBV. (A) Hierarchical clustering analysis and the heat map demonstrate the different levels of IncRNA expression between AGS-EBV and AGS. The colour scale of the strips runs from green (low relative expression) through black (medium relative expression) to red (high relative expression). (B) Volcano plot displaying the differentially expressed IncRNA. The vertical green lines refer to a 1.5-fold (log2 scaled) up-regulation and down-regulation, respectively. The horizontal green line corresponds to a P-value of 0.05 (−log10 scaled). The red points in the plot represent up-regulated mRNAs with statistical significance, while the down-regulated IncRNAs are shown in green. (C) Expression profile of IncRNA isoform. (D) The top 50 enrichment scores in gene ontology (GO) enrichment analysis of dysregulated IncRNA. Red bars represent cell component terms. Green bars represent molecular function terms. (E-H) Top 15 classes of KEGG and Reactome pathway enrichment terms, respectively. The larger the rich factor the greater the enrichment.
functions (Figure s2E-G). This indicates that post-transcriptional protein modification involved in multiple pathways is the main function of DE circRNAs after EBV infection.

**EBV-Encoded circRNA: ebv_circ_RPMS1**

To our knowledge, this is the first time that circRNA has been verified to be widespread in many species by next-generation sequencing, and we speculated whether EBV can encode circular RNA. After comparing EBV-AGS and AGS, we discovered a circRNA derived from the EBV viral gene RPMS1 by sequencing, which has been validated in EBV-associated cell lines and tissues. In addition, this circRNA not only exists in the cytoplasm but also in the nucleus, and overexpressed viral circRNA can promote the proliferation of AGS cells.

**The DE Genes Were Verified by qPCR**

The 12 selected DE genes that may regulate viral infection and cell growth were verified by qRT-PCR, and the
Verification results were compared with the transcriptome sequencing results (Figures 6 and s4). The linear correlation analysis further verified that the expression trend of the selected genes was consistent with transcriptome sequencing.

**Construction of the circRNA-miRNA-mRNA Network**

Current studies on the function of circRNAs as miRNA sponges have elucidated the genetic and epigenetic regulatory mechanisms of various tumours; therefore, we used DE
circRNAs to predict the miRNAs that could bind to them. The circRNA-miRNA-mRNA network was constructed based on the mRNA obtained through further analysis of its linear correlation with circRNA. Based on the prediction of miRNA pairing by the DE circRNA, 22,980 miRNAs were obtained (Figure 7A). The top 100 records were selected for subsequent analysis according to the total score, including 33 circRNAs and 81 miRNAs. The host genes of 33 circRNAs were crossed with the predicted target mRNAs of 81 miRNAs, and 21 overlapping mRNAs were obtained. A total of 91 key molecules involved in the EBVaGC cell line were explored, and the circRNA-miRNA-mRNA network is shown in Figure 7B. The establishment of a regulatory network may provide a new method for clarifying the progression of EBVaGC.

Discussion

Benefiting from the high expression of EBV in a single infected cell (10 million copies per cell) and its monoclonal proliferation characteristics, we successfully constructed a stable and virus-rich EBV-positive gastric cancer cell line by using cell-to-cell infection. Specific molecular expression profiles of host cells after virus infection can be effectively revealed by sequencing, such as changes in the expression levels of coding or non-coding RNA caused by cytomegalovirus, human
immunodeficiency virus, and human papillomavirus infection. Then, function and pathway analyses were used to predict the potential mechanisms of disease progression. 18

In our study, multiple DE mRNAs were changed by nearly 200-fold in AGS-EBV, many of which were oncogenes or tumour suppressor genes. Phospholipase A2 group IIA (PLA2G2A), UDP glucuronosyltransferase family 1 member A6 (UGT1A6), and meiotic recombination protein 8 (REC8) are three of the genes that are closely related to gastric cancer, consistent with our sequencing and verified in other studies. Overexpression of PLA2G2A in gastric cancer is associated with improved survival in patients and enhances AGS invasion through the beta-catenin-dependent Wnt pathway. The qPCR detection of gastric cancer tissue showed that the expression of the oncogene UGT1A6 was up-regulated and the tumour suppressor REC8 was down-regulated, which was more significant in EBVaGC. Bioinformatics analysis further suggests that the role of DE mRNA in EBVaGC may be to facilitate the completion of EBV-infected cells and modulate host cell mitosis. In a typical example, DE mRNA serpin family A member 1 (SERPINA1) was found to be up-regulated by 208-fold in AGS-EBV, and GO enrichment suggested that it was primarily involved in glycoprotein binding. Researchers have shown that glycoprotein is an important factor for EBV to enter B cells, and glycoprotein homeostasis is conducive to regulate host immune defence mechanisms, which affects the infection of host cells with EBV. 19 Additionally, in vitro cell studies have also found that plasmids carrying both the SERPINA1 gene and EBV-encoding gene EBNA1 can significantly promote chromosome replication in mammalian cells. 20 Moreover, the SERPINA1 gene was significantly overexpressed in the serum of gastric cancer patients compared with healthy individuals, which was correlated with clinical staging. 21 This indicated that these DE mRNAs may play a special role in oncogene activation or tumour suppressor gene inactivation and may be potential biomarkers for EBVaGC.

Next, by comparing transcriptional sequencing with the latest complete expression profile of EBVaGC mRNA, we read more EBVaGC virus transcripts. As a further complement to the understanding of EBV encoding mRNA, transcriptome analysis revealed that AGS-EBV expressed 76 EBV genes previously detected in EBVaGC, as well as 15 unreported genes. In contrast to EBV-related diseases such as nasopharyngeal cancer and lymphoma, 4 of the 15 genes found in EBvaGC (RPMS1, LF1, LF2, and LF3) had higher expression levels than LMP2A and EBNA-1. LMP2A is usually one of the most highly expressed transcripts of EBV because most of the viruses in EBVaGC tissue are latent. The main reason for this result was that EBV was sequenced in a short period after AGS infection, with most of the virus still in the lytic state. Consistent

![Figure 7](https://oncotargetsandtherapyjournal.com/article-figures/fig7.png)

**Figure 7** The R software was used to show the circRNA-miRNA-mRNA interactions. (A) mRNA match by the dysregulation circRNA. (B) Construction of the circRNA-miRNA-mRNA network. Total of the top 100 connection RNA-related. Diamond represents circRNA, square represents mRNA, and circle represents mRNA. The most connected molecule indicates the most important in the network of relationships.
with our hypothesis, BDLF3.5 is the lowest read viral gene, which is related to the formation of late viral gene regulatory complexes.22 The four undetected genes, BARF0, BdRF1, BLLF2, and BNLF2a, were also associated with the late viral gene.23 In other words, this study found that highly expressed viral genes may mainly be involved in the early entry of viruses into host cells and the oncogenesis of EBVaGC. During the EBV propagation process, herpesviruses exhibit a strictly regulated temporal cascade of gene expression that can be divided into three general stages: immediate-early, early, and late. This prompts us to refer to the specificity of EBVaGC progression, and the high expression mRNA of EBVaGC may serve as a new research target.

When studying lncRNAs for whole-transcript sequencing, we found a star lncRNA, Lnc-XIST: 7, which was up-regulated 5128 times in AGS-EBV and also showed the greatest fold change of all DE lncRNAs. Lnc-XIST: 7 has been shown to activate the Wnt/β-catenin signalling pathway in a variety of digestive system tumours and promote tumour cell proliferation, lymph node metastasis, and chemotherapy resistance.24 It may also be the next hot spot for the study of lncRNAs in EBVaGC. Bioinformatics analysis of the multiple functions and pathways enriched by DE lncRNAs in AGS-EBV suggests that the progression of EBVaGC is diverse. Among them, lncRNA binding to proteins may be the most important potential mechanism, which is consistent with the microarray data analysis results of 12 EBVaGC tissues in the GEO database.25 In recent years, several lncRNAs (LncClnC1, LncSTCAT3, and LncGMAN) have been confirmed to regulate the invasion of EBVaGC tissues in the GEO database.26 Therefore, further studies are needed to confirm whether lncRNAs can play a role in EBVaGC by binding to actin.

Marquitz has shown that two types of long noncoding RNAs transcribed from the EBV BamHI A Rightward Transcript Region contribute to growth regulation, which explains how EBV can encode viral lncRNA called BART RNAs.28 Relevant lncRNAs were not read out in this sequencing study, and the reason may be that the abundance level of lncRNAs is usually 10 times lower than that of mRNA. In whole transcriptome sequencing, mRNA data volume is relatively large, and viral lncRNAs with fewer reads are easily covered.

EBV infection changes mRNA and lncRNA expression profiles of GC, as well as that of circRNA. To the best of our knowledge, this is the first study on the expression of host circRNAs in EBVaGC. We found that circRNA was widely expressed in EBVaGC, and the length was similar to that of most current circRNA (400–600nt), showing that the conserved characteristic of circRNA was also present in EBVaGC. EBVaGC’s DE circRNA is mainly associated with DNA repair during cell mitosis and protein modification, similar to the GO hypothesis, and the best studied function so far is the miRNA sponge function. The most up-regulated DE circRNA, circNRIP1, has been shown to promote gastric cancer progression by targeting mirRNA-149-5p by stimulating the AKT1/mTOR pathway.29 Another up-regulated DE circRNAs-circCDTP1 also played an oncogenic role in nasopharyngeal carcinoma via the mirRNA-320b/HOXA10/TGFβ2 pathway. Knockdown of circCDTP1 resulted in the inhibition of proliferation, migration, and invasion, and promoted apoptosis of nasopharyngeal carcinoma cells.30

The ebv_circ_RPMS1 comes from the BART region, which is prone to selective splicing to produce a variety of transcripts and multiple non-coding RNAs. After optimising the sequencing process and inducing reactivation of EBV, Nathan and Ungerleider found that other viral genes EBN4, LMP2, BHLF1, and BHRF1 can also produce circRNA in lymphoma and EBVaGC.31,32 According to our previous studies, ebv_circRNA might act as a sponge of miRNA and accelerate the proliferation of EBVaGC cell lines. In Liu et al study,33 knockdown of ebv_circ_RPMS1 inhibited EBV-positive NPC cell proliferation, induced apoptosis, and repressed cell invasion. Investigation of the mechanism also revealed that ebv_circ_RPMS1 mediated NPC oncogenesis by sponging multiple miRNAs and promoting epithelial-mesenchymal transition. The inhibitors of miR-203, miR-31, and miR-451 could reverse the effects of ebv_circ_RPMS1 knockdown on NPC cells. Additionally, our lab found that another
circRNA encoded by the EBV named ebv_circ_LMP2A was highly expressed in EBVaGC tissues and plays crucial roles in inducing and maintaining GC stemness phenotypes by targeting the miR-3908/TRIM59/p53 axis. Furthermore, high expression of ebv_circ_LMP2A is significantly associated with metastasis and poor prognosis in patients with EBVaGC.

In the study of the EBVaGC progression mechanism, we speculated that there might be a regulatory mechanism of circRNAs, miRNAs, and mRNAs involved in the occurrence of EBVaGC. By constructing a circRNA-miRNA-mRNA network, we screened nuclear FMR1 interacting protein 2 (NUFIP2), and miR-190a-3p, circ_0006251 (circTET3), zinc finger, BTB domain containing 7A (ZBTB7A), and miR-103a-3p were the five most connected molecules located at the centres of the network. NUFIP2 emerges as a cofactor that contributes to mRNA target recognition by Roquin, which is essential for appropriate immune cell function. Similar to Hamzeiy et al, NUFIP2 serves as a target gene for multiple miRNAs associated with EBVaGC. Considerable evidence exists to suggest that miR-190a-3p and miR-103a-3p significantly enhanced the proliferation, migration, and invasion abilities of GC cell lines. High expression of ZBTB7A was significantly correlated with poorer prognosis in GC patients, and miR-100 has been shown to reduce GC aggressiveness by directly targeting ZBTB7A. In addition, circTET3 can mediate migration of rat vascular smooth muscle cells by targeting miR-351-5p. These lines of evidence indicated that the molecules in this network are highly correlated with GC, and both play a role by targeting miRNA. Intriguingly, most of them showed distinct differences in expression after EBV infection. The circRNA-miRNA-mRNA network is expected to shed light on the core elements of EBVaGC progression.

Conclusion

In general, this research integrated multilevel expression data and a bioinformatics approach to reveal key genes involved in the mechanism of EBVaGC cell lines and their network regulation. Our RNA-sequencing results will help expand our understanding of EBVaGC and provide an indispensable resource for related research on genomics, genetics, and molecular mechanisms in the future.

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Disclosure

The authors report no conflicts of interest in this work.

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