Identification and prediction of novel non-coding and coding RNA-associated competing endogenous RNA networks in colorectal cancer

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Abstract

AIM
To identify and predict the competing endogenous RNA (ceRNA) networks in colorectal cancer (CRC) by bioinformatics analysis.

METHODS
In the present study, we obtained CRC tissue and normal tissue gene expression profiles from The Cancer Genome Atlas project. Differentially expressed (DE) genes (DEGs) were identified. Then, upregulated and downregulated miRNA-centered ceRNA networks were constructed by analyzing the DEGs using multiple bioinformatics approaches. DEMRNAs in the ceRNA networks were identified in Kyoto Encyclopedia of Genes and Genomes (KEGG) pathways using KEGG Orthology Based Annotation System 3.0. The interactions between proteins were analyzed using the STRING database. Kaplan-Meier survival analysis was conducted for DEGs and real time quantitative polymerase chain reaction (RT-qPCR) was also performed to validate the prognosis-associated lncRNAs in CRC cell lines.

RESULTS
Eighty-one DElncRNAs, 20 DEMiRNAs, and 54 DEMRNAs were identified to construct the ceRNA networks of CRC. The KEGG pathway analysis indicated that nine out of top ten pathways were related with cancer and the most significant pathway was "colorectal cancer". Kaplan-Meier survival analysis showed that the overall survival was positively associated with five DEGs (IGF2-AS, POU6F2-
AS2, hsa-miR-32, hsa-miR-141, and SERPINE1) and it was negatively related to three DEGs (LINC00488, hsa-miR-375, and PHLPP2). Based on the STRING protein database, it was found that SERPINE1 and PHLPP2 interact with AKT1. Besides, SERPINE1 can interact with VEGFA, VTN, TGFBI, PLAU, PLAUR, PLG, and PLAT. PHLPP2 can interact with AKT2 and AKT3. RT-qPCR revealed that the expression of IGF2-AS, POU6F2-AS2, and LINC00488 in CRC cell lines was consistent with the in silico results.

CONCLUSION
CeRNA networks play an important role in CRC. Multiple DEGs are related with clinical prognosis, suggesting that they may be potential targets in tumor diagnosis and treatment.

Key words: Colorectal cancer; LncRNA; MicroRNA; Overall survival; Competing endogenous RNA; Bioinformatics analysis

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INTRODUCTION
Colorectal cancer (CRC) is one of the most common digestive malignancies in the world[1]. With the development of new technology in CRC diagnosis and treatment, the prognosis of patients with early detection has been improved. However, the overall 5-year survival rate in advanced cases remains poor[2]. In the United States, CRC was the fourth most common malignant tumor, with 135,430 new cases and 50,260 deaths in 2017[3]. There-}

...n the prognosis of patients with early detection has been improved. However, the overall 5-year survival rate in advanced cases remains poor[2]. In the United States, CRC was the fourth most common malignant tumor, with 135,430 new cases and 50,260 deaths in 2017[3]. Therefore, specific CRC biomarkers and therapeutic pathways are in great need to improve the prognosis for patients.

Non-coding RNAs, including microRNAs (miRNAs) and long-noncoding RNAs (lncRNAs), can regulate oncogene and tumor suppressor gene expression in multiple ways[4,5]. miRNAs are 20–22 nucleotides long and regulate genes post-transcriptionally by directly binding miRNAs[6]. LncRNAs are defined as transcripts that range from 200 nucleotides to multiple kilobases in length[7]. Recent research has focused on these lncRNAs, which function as competing endogenous RNAs (ceRNAs) to regulate gene expression by sponging miRNAs through shared miRNA response elements[8].

In the last few years, with the development of gene-sequencing technology, dysregulation of lncRNAs has been revealed in diverse malignancies. Studies have utilized bioinformatics tools to predict the target genes of novel lncRNAs, and molecular biology techniques including real time quantitative polymerase chain reaction (RT-qPCR), silencing technique, and luciferase reporter gene assays, among others, to validate in silico predictions. In CRC, the lncRNA UICLM acts as a ceRNA for hsa-miR-215 to upregulate ZEB2 expression and promote CRC progression[9]. Additionally, the lncRNA HNF1A-AS1 functions as an oncogene in the metastasis of CRC by modulating the hsa-miR-34/p53 axis[10]. Finally, the lncRNA CAS2 plays a role as a tumor suppressor gene by sponging hsa-miR-18a[11].

The lncRNA/miRNA/mRNA axis is regarded as an important mechanism in tumor progression and metastasis[12]. However, studies of ceRNA networks of novel coding and noncoding RNAs in CRC in large cohorts have not been performed. In our study, we obtained malignant and normal tissue expression profiles from The Cancer Genome Atlas (TCGA, http://cancergenome.nih.gov) project[13]. CeRNA networks were constructed including differentially expressed miRNAs (DEmiRNAs), differentially expressed lncRNAs (DElncRNAs), and differentially expressed mRNAs (DEmRNAs), which were based on the miRcode (http://www.mircode.org/)[14], miRTarBase (http://mirbase.mbc.nctu.edu.tw/php/) [15], TargetScan (http://www.targetscan.org/), and miRDB (http://www.mirdb.org/) databases[16]. Kaplan-Meier survival curve analysis was performed to identify differentially expressed genes (DEGs) that are associated with overall survival.

MATERIALS AND METHODS
Sample collection
We downloaded CRC transcriptome profiles from TCGA through the Genomic Data Commons (GDC) Data Transfer Tool 1.3.0[13]. The public data included the tissue expression profiles (derived by RNA-seq of 644 CRC tissues and 51 normal tissues (level 3) and 619 CRC and 11 normal tissue expression profiles (level 3) derived by miRNA-seq. According to the publication guidelines (2015) provided by TCGA (https://cancergenome.nih.gov/publications/publicationguidelines), our study does not require the approval of an ethics committee.

Identification of DEGs
We analyzed the RNA-seq data by merging it to an RNA matrix with PERL software. Then, we converted the gene
ID to gene name according to Ensembl (Homo sapiens) (http://asia.ensembl.org/index.html). The miRNA-seq data were analyzed using the same method. DEmRNAs, DElncRNAs, and DEmiRNAs were identified with the edgeR package in R with a threshold log2 fold change (FC) > 2.0 and P < 0.01. Heat maps of DEGs were constructed using the gplots package in R.

Functional analysis and ceRNA network construction

Next, we constructed ceRNA networks that were mapped by identifying DEGs to an established co-expression database (miRcode, miRTarBase, TargetScan, and miRDB). LncRNA-miRNA-mRNA reactions were split into IncRNA-miRNA and miRNA-mRNA interactions. LncRNA-miRNA interactions were predicted by comparing DElncRNAs and DEmiRNAs to the miRcode database. DEmiRNAs regulating genes of DEmiRNAs were identified by intersecting the predicted results between miRTarBase, TargetScan, and the miRDB databases. According to the ceRNA hypothesis\[^{12}\], miRNAs negatively regulate mRNA expression and lncRNAs act as ceRNAs to limit mRNA function by sponging them. The networks were visualized and mapped using Cytoscape v3.5.1\[^{17}\].

To explore the function of the DEGs in the ceRNA networks in tumorigenesis and metastasis, KEGG pathways\[^{18}\] were analyzed using KOBASE (http://kobas.cbi.pku.edu.cn/), which is a web server that annotates an input set of genes with pathways based on human disease databases\[^{19}\]. Then, we utilized STRING protein database v10.5 (http://string-db.org/) to analyze the protein-protein interactions of the DEGs.

Association analysis between DEGs and CRC patient survival

We selected CRC adenocarcinoma and adjacent normal tissue expression data and clinical information from TCGA project. None of the patients received preoperative treatment. Kaplan-Meier survival analysis and the log-rank test were performed for DEGs in the ceRNA networks. P < 0.05 was regarded as statistically significant.

Cell lines and cell culture

The CRC cell lines HT29, LoVo, and SW480, along with the normal intestinal epithelial cell line NCM460, were purchased from the Institute of Biochemistry and Cell Biology at the Chinese Academy of Sciences, Shanghai, China. The cell lines were cultured in RPMI 1640 medium (HyClone, Logan, UT, United States) supplemented with 10% fetal bovine serum (Gibco, Grand Island, NY, United States), 100 U/mL penicillin, and 100 μg/mL streptomycin. All cells were maintained in an incubator with a humidified atmosphere of 95% air and 5% CO₂. All cell lines were maintained in an incubator with 10% fetal bovine serum (Gibco, Grand Island, NY, United States) supplemented with 10% fetal bovine serum (Gibco, Grand Island, NY, United States) and 100 U/mL penicillin, and 100 μg/mL streptomycin. All cells were maintained in an incubator with a humidified atmosphere of 95% air and 5% CO₂ at 37 °C.

DElncRNA detection

Total RNA was isolated from the cells using Trizol reagent (Invitrogen, Carlsbad, CA, United States). The quality and concentration of isolated RNA were detected with a NanoDrop 2000 spectrophotometer. Total RNA (800 ng) was converted to cDNA by using the PrimeScript RT kit (Takara, Otsu, Shiga, Japan). The kit contained gDNA Eraser which could remove genomic DNA. We utilized LightCycler96 (Roche Diagnostic, Basel, Switzerland) to perform RT-qPCR using SYBR Green (Takara). The two-step amplification reaction was as follows: 95 °C for 30 s for preincubation, then 45 cycles of 95 °C for 10 s and 60 °C for 30 s. GAPDH expression was used as an endogenous control. Quantitative analysis was calculated using the 2⁻ΔΔCt method\[^{20}\]. Primers were designed as follows: POU6F2-AS2 forward, 5'–ACAGCAGTCGCCAGAAGGATTTG–3' and reverse, 5'–GCAGACCTAGCTTTGAGTGAC–3'; LINCO0488 forward, 5'–GAGCAGCAAGAATGAGCAGAGG–3' and reverse, 5'–GAATCTGGAAGCACCCTGAACC–3'; IGF2-AS forward, 5’–TCAACACAGACAGACAGACC–3' and reverse, 5’–TCCGTTGGTGGCTCAGGTTG–3'; GAPDH forward, 5'–AGCCACATCGCTCAAGCAC–3' and reverse, 5'–GCCAATACAGCAGAAATCC–3'.

Statistical analysis

All data were analyzed using SPSS 20.0 (Chicago, IL, United States). Student's t-test was performed for CRC cell lines and normal intestinal epithelial cell line comparisons. Differences with P < 0.05 were regarded as statistically significant. All experiments were repeated three times.

RESULTS

Identification of DEGs in CRC

We identified DElncRNAs and DEmRNAs between 644 CRC and 51 normal tissue expression profiles from TCGA. As a result, we identified 1043 DElncRNAs, 2146 DEmRNAs, and 276 DEmiRNAs using the edgeR package in R. There were 768 up-regulated and 275 down-regulated DEGs among the DElncRNAs, and 1198 up-regulated and 948 down-regulated DEmiRNAs in the DEmiRNAs. Using the same method, 180 up-regulated and 96 down-regulated DEmiRNAs were obtained by comparing 619 CRC and 11 normal tissue expression profiles. Moreover, we constructed heat maps of the top DEGs (log2 FC > 5, P < 0.01) in each category using R (Figures 1-3).

Construction of ceRNA networks

To construct the ceRNA networks, we predicted the interactions between DElncRNAs and DEmiRNAs and we also predicted DEmiRNAs targeted by the DEmiRNAs by intersecting the predictions of miRTarBase, TargetScan, and miRDB (Figure 4). Ultimately, 81 DElncRNAs, 20 DEmiRNAs, and 54 DEmiRNAs were identified to construct the ceRNA networks of CRC using Cytoscape (Figure 5). The interactions among DEGs are shown in Tables 1 and 2.

Functional analysis of DEmiRNAs in ceRNA networks

To better reveal the function of ceRNA networks in CRC, we subjected the 54 DEmiRNAs in the ceRNA network to Kyoto Encyclopedia of Genes and Genomes...
Figure 1  Heat map of differentially expressed long-noncoding RNAs (log: FC > 5, P < 0.01). The above horizontal axis shows clusters of samples. The left vertical axis shows clusters of differentially expressed long-noncoding RNAs (DElncRNAs) and right vertical axis represents lncRNA names. Red represents up-regulated genes and green represents down-regulated genes. The yellow color column represents colorectal cancer; the blue represents normal.
Figure 2 Heat map of differentially expressed mRNAs (log: FC > 5, \(P < 0.01\)). The above horizontal axis shows clusters of samples. The left vertical axis shows clusters of differentially expressed mRNAs and right vertical axis represents mRNA names. Red represents up-regulated genes and green represents down-regulated genes. The yellow color column represents colorectal cancer; the blue represents normal.
Figure 3  Heat map of differentially expressed microRNAs (log2 FC > 5, P < 0.01). The above horizontal axis shows clusters of samples. The left vertical axis shows clusters of differentially expressed microRNAs (DEmiRNAs) and right vertical axis represents miRNA names. Red represents up-regulated genes and green represents down-regulated genes. The yellow color column represents colorectal cancer; the blue represents normal.
(KEGG) pathway analysis using KEGG Orthology Based Annotation System (KOBASE) 3.0, and the top ten KEGG pathways are shown in Table 3. There were nine pathways related to cancer, including "Colorectal cancer", "p53 signaling pathway", "Chronic myeloid leukemia", "Proteoglycans in cancer", "miRNAs in cancer", "Thyroid cancer", "Transcriptional misregulation in cancer", "Pathways in cancer", and "Wnt signaling pathway". These results indicate that ceRNA networks play important roles in the carcinogenesis and progression of CRC.

**DEGs related with prognosis of CRC patients**

We performed Kaplan-Meier curve analysis to identify the DEGs in the ceRNA networks that are related to overall survival. The results showed that 3 DElncRNAs (IGF2-AS, POU6F2-AS2, and LINCO0488), 3 DEMiRNAs (hsa-miR-32, hsa-miR-141, and hsa-miR-375), and 2 DEMiRNAs (PHLP2 and SERPINE1) were associated with the clinical prognosis of patients. The Kaplan-Meier curves showed that the IncRNAs IGF2-AS and POU6F2-AS2, as well as hsa-miR-32, hsa-miR-141, and serpin peptidase inhibitor, clade E member 1 (SERPINE1) were negatively correlated with overall survival, whereas LINCO0488, hsa-miR-375, and PH domain and leucine rich repeat protein phosphatase 2 (PHLP2) were positively correlated with overall survival (Figure 6).

The ceRNA networks illustrated that PHLP2 may be regulated by hsa-miR-32 and hsa-miR-141. Interestingly, PHLP2 as well as hsa-miR-32 and hsa-miR-141 has been previously related to patient prognosis. We used
Table 2  Putative differentially expressed microRNAs that may target differentially expressed mRNAs

| DEmiRNAs       | DEmRNAs                     |
|----------------|-----------------------------|
| Hsa-miR-193b   | PLAU, PMAIP1, TCF7, CCND1, SHMT2 |
| Hsa-miR-145   | SERPINE1, SOX2, MUC1, YES1   |
| Hsa-miR-187   | LYN, RNPS1, MYLIP           |
| Hsa-miR-375   | TRIM66, USP1, TGFβ2, COL12A1, CBX3, SPI |
| Hsa-miR-150   | CBL, ZEB1, HILPDA, SLC7A11, DACH1, EREG, IGF2BP3, KIAA1549, MYB |
| Hsa-miR-183   | KIF5C                       |
| Hsa-miR-98    | HAND1                       |
| Hsa-miR-144   | CRKI3                       |
| Hsa-miR-182   | CHL1, FOXF2, TCEAL7, NPTX1  |
| Hsa-miR-152   | NPTX1, BMP3, KLF4           |
| Hsa-miR-454   | RIM20, CFL2                 |
| Hsa-miR-106a  | CFL2, FAM129A, CADM2        |
| Hsa-miR-372   | TMEM100, CADM2, SLC16A9, CYBRD1, KLF4, CADM2, FAM129A |
| Hsa-miR-21    | ATP2B4, EDL13, OSR1         |
| Hsa-miR-32    | UGP2, PBLD, PHLP2, ATP2B4  |
| Hsa-miR-141   | ELAVL4, EPHA7, PHLP2        |
| Hsa-miR-338   | NOVA1                       |
| Hsa-miR-206   | SFRP1                       |

Table 3  Top 10 Kyoto Encyclopedia of Genes and Genomes pathways enriched by the differentially expressed mRNAs involved in competing endogenous RNAs networks

| Pathway ID  | Description                          | P value | Gene name                     |
|-------------|--------------------------------------|---------|-------------------------------|
| hsa05210    | Colorectal cancer                    | 8.18E-05| TCF7, CCND1, TGFB2           |
| hsa04115    | p53 signaling pathway                | 1.11E-04| SERPINE1, CCND1               |
| hsa05220    | Chronic myeloid leukemia             | 1.30E-04| CCND1, CBL, TGFβ2            |
| hsa02050    | Proteoglycans in cancer              | 1.52E-04| CCND1, CBL, TGFβ2, PLAU      |
| hsa05206    | MicroRNAs in cancer                  | 6.20E-04| ZEB1, CCND1, PLAU, TGFβ2     |
| hsa05216    | Thyroid cancer                       | 7.31E-04| TCF7, CCND1                  |
| hsa04550    | Signaling pathways regulating pluripotency of stem cells | 8.67E-04| SOX2, KLF4, HAND1           |
| hsa04310    | Wnt signaling pathway                | 8.84E-04| TCF7, CCND1, SFRP1           |
| hsa05202    | Transcriptional misregulation in cancer | 1.69E-03| SPI, PLAU, ZEB1              |
| hsa05200    | Pathways in cancer                   | 1.74E-03| TCF7, CCND1, CBL, TGFβ2      |

Figure 5  CeRNA networks of long-noncoding RNA-microRNA-mRNA in colorectal cancer. The red represents the upregulated and the blue represents downregulated. Diamonds represent long-noncoding RNAs, balls represent mRNAs, and rectangles represent microRNAs.
the STRING protein database to analyze the protein-protein interactions of SERPINE1 and PHLPP2. The results revealed that SERPINE1 and PHLPP2 interact with V-akt murine thymoma viral oncogene homolog 1 (AKT1). Moreover, SERPINE1 can interact with several proteins including vascular endothelial growth factor A (VEGFA), vitronectin (VTN), transforming growth factor, beta 1 (TGFβ1), plasminogen activator, urokinase (PLAU), plasminogen activator, urokinase receptor (PLAUR), plasminogen (PLG), and plasminogen activator, tissue (PLAT). Additionally, PHLPP2 can interact with AKT2 and AKT3 (Figure 7).

**Expression of IGF2-AS, POU6F2-AS2, and LINC00488 in cell lines**
To further confirm our *in silico* results, we performed RT-qPCR to detect the expression levels of IGF2-AS, POU6F2-AS2, and LINC00488 in CRC cell lines (HT29, LoVo, and SW480) and in a normal intestinal epithelial cell line (NCM460). We found that the expression of IGF2-AS and POU6F2-AS2 was significantly increased in CRC cell lines. Conversely, LINC00488 was downregulated in CRC cell lines compared with NCM460 (Figure 8).

**DISCUSSION**

The role of ceRNAs in tumorigenesis and development has remained controversial[21]. However, many studies have revealed the regulatory function of ceRNA networks in proliferation, invasion, metastasis, and epithelial-mesenchymal transition (EMT). This urgently requires the establishment of a comprehensive regulatory net-
work based on a large sample size of whole genome sequences. By constructing regulatory networks, we can better clarify the role of ceRNAs and provide direction for further research.

In this study, we integrated IncRNA, miRNA, and mRNA high-throughput data from TCGA and constructed ceRNA networks. KEGG pathway analysis of DEmRNAs indicated that ceRNA networks may regulate CRC progression via multiple mechanisms. It is noteworthy that "Colorectal cancer" was the most significant pathway in which transcription factor 7 (TCF7), cyclin D1 (CCND1), and transforming growth factor, beta 2 (TGFB2) are involved. TGF7 is a critical signaling molecule in the WNT/j-catenin pathway, which belongs to the TCF/LEF1 family. The WNT/j-catenin pathway is recognized as a key regulator in cancer by transcriptionally activating a variety of oncogenes including CCND1. CCND1 has been validated as an oncogene in CRC and it regulates the cell cycle transition from the G1 phase to the S phase. Chen et al. showed that TGFβ2 induces CRC migration, metastasis, and the EMT by promoting SNAIL and SLUG expression.

To identify novel prognosis-related biomarkers, we applied Kaplan-Meier curve analysis to identify DEGs in ceRNA networks that correlated with clinical features among CRC patients. Ultimately, we detected 3 DEmiRNAs (hsa-miR-32, hsa-miR-141, and hsa-miR-375), and 2 DEmRNAs (PHLPP2 and SERPINE1) that may be indicators of prognosis. Wu et al. showed that hsa-miR-32 promoted CRC growth, migration, and invasion by downregulating PTEN. Feng et al. revealed that hsa-miR-141 may act as a biomarker for the early detection of recurrence during CRC surveillance. Cui et al. found that hsa-miR-375 acts as an anti-oncogene through the inhibition of SP-1, BCL-2, and other EMT-associated genes. Increased SERPINE1 expression has been found in G3/G4 CRC cells, which may be a predictor of CRC invasiveness, progression, and overall survival. PHLPP2, a protein phosphatase, is an isoform of PHLPP. PHLPP2 negatively regulates RAF/MEK/ERK signaling by directly inhibiting RAF1 activity to inhibit the progression of cancer. Li et al. and Liao et al. reported that hsa-miR-938 and hsa-miR-224 repressed PHLPP2 expression in CRC. In the ceRNA networks identified here, our prediction showed that PHLPP2 is targeted by hsa-miR-141, hsa-miR-32, and hsa-miR-424. We noted that the hsa-miR-141/PHLPP2 and hsa-miR-32/PHLPP2 axis in ceRNA networks may be a diagnostic biomarker and a therapeutic target for the treatment of CRC.

In the present study, 1043 DEmiRNAs were detected in 644 CRC tissues compared to 51 normal tissues. Among them, 3 DEmiRNAs (IGF2-AS, POU6F2-AS2, and LINC00488) were related to the prognosis of CRC. The IncRNA IGF2-AS is an antisense IncRNA for IGF2, and has been validated to promote hepatitis C virus replication. In our work, IGF2-AS may regulate target genes by competitive sponging against hsa-miR-150 and hsa-miR-193b. The IncRNA POU6F2-AS2 has been demonstrated to be overexpressed in esophageal squamous cell cancer, which can directly target the Ybx protein and protect cancer cells from ionizing radiation. We predicted that hsa-miR-375 interacts with POU6F2-AS2, which is related with clinical prognosis. Additionally, hsa-miR-375 is related to overall survival. The POU6F2-AS2/hsa-miR-375 axis may be another crucial diagnosis-related target. However, little is known about LINC00488. To confirm the expression levels of these three IncRNAs, we performed RT-qPCR in CRC cell lines. The results were consistent with the in silico analysis results. Additionally, little is known about the functions of IGF2-AS, POU6F2-AS2, and LINC00488 in CRC. Thus, additional research is needed to explore the biological and molecular mechanisms of these DEmiRNAs in CRC.

In summary, we analyzed the expression profiles of CRC samples from TCGA and constructed ceRNA networks of the DEGs. KEGG pathway analysis further confirmed the role of these ceRNA networks in the development of CRC. Moreover, we identified several DEGs that were related to clinical prognosis, and the expression of IGF2-AS, POU6F2-AS2, and LINC00488 was validated using RT-qPCR in cell lines. Our study deepens our understanding of ceRNA networks and provides potential therapeutic targets and prognosis-related biomarkers for further research.

**ARTICLE HIGHLIGHTS**

**Research background**

With the development of high-throughput technology, dysregulation of non-coding genes has been revealed in colorectal cancer (CRC). Furthermore, accumulating studies have demonstrated that long-noncoding RNAs (IncRNAs) function as competing endogenous RNAs (ceRNAs) to regulate oncogene and tumor suppressor gene expression by sponging microRNAs (miRNAs). In the present research, we constructed and analyzed the ceRNA networks and found the prognosis-related differentially expressed genes (DEGs) by bioinformatics.
The expression of the three long-noncoding RNAs in HT29, LoVo, and SW480 and normal intestinal epithelial cell line NCM460 was detected by RT-qPCR and GAPDH was used as an internal control. \( P < 0.01 \).

**Figure 8** IGF2-AS, POU6F2-AS2, and LINC00488 expression levels in colorectal cancer cell lines. The expression of the three long-noncoding RNAs in HT29, LoVo, and SW480 and normal intestinal epithelial cell line NCM460 was detected by RT-qPCR and GAPDH was used as an internal control. \( P < 0.01 \).
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