A two-circular RNA signature of donor circFOXN2 and circNECTIN3 predicts early allograft dysfunction after liver transplantation

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Background: Early allograft dysfunction (EAD) following liver transplantation is associated with poor recipient and graft survival. In recent years, circular RNAs (circRNAs) have emerged as important components of endogenous RNAs. This study aims to explore the expression profile and predictive value of graft circular RNAs for EAD after liver transplantation.

Methods: RNA sequencing was conducted to identify the circRNA profile in donor liver tissues. Furthermore, quantitative real-time polymerase chain reaction (qRT-PCR) was used to identify candidate circRNAs. A novel model combining circular RNA signature was established to predict EAD based on the multivariate analysis.

Results: A total of 442 circRNAs were differentially expressed between the EAD and non-EAD groups, of which, 223 were significantly upregulated and 219 were downregulated in the EAD group (Fold change >2, P<0.05). qRT-PCR validation indicated that circFOXN2 and circNECTIN3 levels in the EAD group were significantly lower than those in the non-EAD group (P=0.038, 0.024, respectively; n=115). Among the 115 recipients, 32 recipients with high circFOXN2 expression were classified as circular RNA signature A and the rest recipients with low circFOXN2 expression were categorized into circular RNA signature B (n=33, high circNECTIN3 expression) and C (n=50, low circNECTIN3 expression). The incidence rates of EAD in signature A, B and C were significantly different (3.1%, 21.2% and 42.0%, respectively; P=0.000). According to the multivariate analysis, a novel predictive model for EAD was developed based on CIT (P=0.000) and circular RNA signature (P=0.013). The novel model displayed a high predictive value for EAD with areas under the curve (AUC) of 0.870 (95% CI: 0.797–0.942).

Conclusions: Donor circFOXN2 and circNEXTIN3 were associated with the incidence of EAD. The novel model combing the two-circular RNA signature had a high predictive value for EAD.

Keywords: Circular RNA; early allograft dysfunction (EAD); liver transplantation

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Introduction

Liver transplantation is considered to be the radical curative therapy for end-stage liver disease (1,2). Although perioperative management and surgical techniques have greatly improved, there are still some complications affecting the prognosis of recipients following liver transplantation. The incidence of early allograft dysfunction (EAD) following liver transplantation varies from 20% to 40% according to the liver transplantation centre (3-5). EAD significantly affects the survival rates of both grafts and recipients. It is important to achieve early prediction of EAD. Several studies have attempted to predict EAD based on clinical parameters, such as the recipient Model for End-stage Liver Disease (MELD) score or cold ischemia time (CIT) (6,7). In addition, microRNAs (miRNAs) and genes in donor liver tissues have been used to predict EAD (8,9).

Circular RNAs (circRNAs) are special endogenous RNAs with circular structures. The formation of circRNAs mainly goes through four spliceosomedependent, intron-pairing-driven, lariat-driven and protein factor-associated circularization paths (10). CircRNAs play a key role in regulating gene expression by acting as a sponge for miRNAs (11). Due to the circular structure, circRNAs are often more stable than linear RNA molecules, which make the former a perfect candidate as predictive biomarkers. A previous study has shown that circRNAs have good predictive value for lung adenocarcinoma (12). Other researchers have also found the predictive value of circRNA in patients with active tuberculosis (13).

However, the expression profile and predictive value of graft circular RNAs in EAD are still unclear. In the present study, we aim to investigate the expression profile of circular RNAs in donor liver tissue and its predictive value for EAD after liver transplantation.

Methods

Study population and data collection

A total of 115 liver transplantation recipients between January 2015 and December 2017 at the First Affiliated Hospital, Zhejiang University School of Medicine, were included in this study. This study was approved by the hospital ethics committee and complied with the World Medical Association's Declaration of Helsinki revised in 2013. Informed consent was obtained from all patients. The

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exclusion criteria included recipients younger than 18 years, recipients undergoing multiorgan transplantation or retransplantation, and recipients with missing essential data for analysis. Donor liver tissue was collected and frozen in liquid nitrogen before storage. All specimens were stored at -80 °C.

EAD was defined by one or more of the following parameters in the revised criteria: (I) total bilirubin (TB) level ≥ 10 mg/dL on postoperative day 7; (II) international normalized ratio (INR) ≥ 1.6 on postoperative day 7; (III) alanine aminotransferase (ALT) or aspartate aminotransferase (AST) levels >3,000 IU/L within the first 7 postoperative days (3,14). The following clinical data were collected: recipient data including age, sex, preoperative MELD score, and operative time, and donor data including donor age, sex, body mass index (BMI), donor liver macrovesicular steatosis and CIT.

CircRNA sequencing

Total RNA was extracted from donor liver tissues using TRIzol reagent (Invitrogen) and treated with DNase I (DNA free kit, Ambion). The RiboMinus Eukaryote Kit (Qiagen) was used to remove ribosomal RNA, and then the samples were treated with RNase R (Epicenter). RNA sequencing (RNA-seq) libraries were prepared using the NEBNext[®] UltraTM RNA Library Prep Kit and sequenced using the Illumina HiSeq 3000 platform at RiboBio Co. Ltd., Guangzhou, China (15,16).

Quantitative real-time PCR (qRT-PCR) and Sanger sequencing

Total RNA was isolated from donor liver tissues using TRIzol reagent (Invitrogen). The concentration and purity of RNA were measured using NanoDrop ND-1000 (Thermo Fisher Scientific, Inc., USA). Reverse transcription and qPCR were carried out using a Geneseed[®] II First Strand cDNA Synthesis Kit and qPCR SYBR[®] Green Master Mix (Guangzhou Geneseed Biotech Co., China), respectively, according to the manufacturer's instructions on an ABI 7500 Fast Real-Time PCR System (Applied Biosystems, CA, USA). The primers were synthesized from Geneseed Biotech Co. (Guangzhou, China). The data were analysed by the 2^{-ΔΔCT} method (17). GAPDH was used as the internal control. The sequences of primers were listed in *Table S1*. The qRT-PCR products of circRNAs were inserted into the T vector and analysed

using Sanger sequencing.

Functional analysis and ceRNA network construction

Kyoto Encyclopaedia of Genes and Genomes (KEGG, https://www.kegg.jp) and Gene Ontology (GO, http:// geneontology.org) were employed to analyse the host genes of dysregulated circRNAs. Potential miRNAs targets of circRNAs were predicted using RNAhybrid and TargetScan. In addition, potential interactions between the miRNAs and genes were predicted with miRWalk and TargetScan. The circRNA-miRNA-mRNA network was constructed using Cytoscape.

Statistical analysis

Continuous parameters in the study were presented as the median and interquartile range or mean \pm SEM. To compare different groups, Student's *t*-test and the Mann-Whitney *U* test were performed where appropriate. Multivariate analysis included variables with a P value <0.05 using a binary logistic regression method. Youden Index was used to select the best cut-off value for EAD. P<0.05 was considered statistically significant. Statistical analyses were performed using GraphPad Prism version 7 (GraphPad Software, Inc. La Jolla, CA, USA) and SPSS version 21.0 (Chicago, IL, USA).

Results

Expression profiles of circRNAs in the EAD and non-EAD groups

As shown in the heat map (*Figure 1A*) and volcano plot (*Figure 1B*), 442 differentially expressed circRNAs were observed (fold change >2, P<0.05) in 3 pairs of donor liver tissue between the EAD and non-EAD groups. Among these circRNAs, 219 were downregulated in the EAD group, and 223 were upregulated. The chromosomal distribution of differentially expressed circRNAs is illustrated in *Figure 1C*. The top 3 Gene Ontology (GO) terms included binding, protein binding and catalytic activity among molecular functions; intracellular part, organelle and intracellular among cellular components; cellular process, metabolic process and organic substance metabolic process among biological processes (*Figure 2A*). The top 3 KEGG pathways related to host genes of the dysregulated circRNAs included arginine biosynthesis, tryptophan metabolism and glyoxylate and dicarboxylate metabolism (*Figure 2B*).

Significantly downregulated circFOXN2 and circNECTIN3 in EAD

According to the distribution and degree of difference based on raw data analysis, circFOXN2 and circNECTIN3 were selected for further research. Then, qRT-PCR was used to detect the expression levels of circFOXN2 and circNECTIN3 in 115 donor liver tissues. The backspliced junctions for circFOXN2 and circNECTIN3 in qRT-PCR products were confirmed by Sanger sequencing (*Figure 3A*). CircFOXN2 and circNECTIN3 were significantly downregulated in the EAD group (circFOXN2: P=0.038, *Figure 3B*; circNECTIN3: P=0.024, *Figure 3C*).

A two-circRNA signature combing donor circFOXN2 and circNECTIN3

According to the Youden index, the cut-off value was set as for 1.27 circFOXN2 (log) and 1.66 for circNECTIN3 (log). We further construct circular RNA signature (A, B and C) according to the cut-off value of circFOXN2 and circNECTIN3. Among the 115 recipients, 32 recipients with high circFOXN2 expression were classified as circular RNA signature A. Recipients with low circFOXN2 expression were categorized into circular RNA signature B (n=33, high circNECTIN3 expression) and C (n=50, low circNECTIN3 expression). The incidence rates of EAD in signature A, B and C were 3.1%, 21.2% and 42.0%, respectively (*Figure 3D*).

A novel predictive model for EAD

The results of univariate analysis of risk factors for EAD are demonstrated in *Table 1*. Preoperative MELD score (P=0.000), CIT (P=0.000) and circular RNA signature (P=0.000) were significantly different between the EAD and non-EAD groups. Furthermore, CIT and circular RNA signature were independent risk factors for EAD based on multivariate binary logistic regression analysis (P=0.000, P=0.013, respectively, *Table 2*). A novel model combing circFOXN2 and circNECTIN3 was constructed according to the independent risk factors. Predictive model score = -7.390 + 0.404*CIT + 2.318*(Signature B) + 3.111*(Signature C). As is shown in *Figure 4*, the novel model showed an improved predictive value with an AUC

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Figure 1 Expression profiles of circRNAs in donor liver tissues. (A) Heat map of differentially expressed circRNAs between the EAD and non-EAD groups; (B) volcano plot of differentially expressed circRNAs between the EAD and non-EAD group; (C) chromosomal distribution of the differentially expressed circRNAs.

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Figure 2 GO (A) and KEGG (B) analyses of host genes of the differentially expressed circRNAs between the EAD and non-EAD groups.



Figure 3 Two candidate circRNAs associated with EAD. (A) Sanger sequencing confirmed the qRT-PCR results of circFOXN2 and circNECTIN3; (B) CircFOXN2 was downregulated in the EAD group compared with that in the non-EAD group (P=0.038); (C) CircNECTIN3 was downregulated in the EAD group compared with that in the non-EAD group (P=0.024); (D) scattergram of circFOXN2 and circNECTIN3 expression (log) identified circular RNA signature A, B and C.

of 0.870 (95% CI: 0.797-0.942).

CircFOXN2/circNECTIN3-miR-135b-5p/miR-149-5p network construction

RNAhybrid and TargetScan were employed to predict the target miRNAs of circFOXN2 and circNECTIN3. To identify the potential targets related to EAD, we selected miRNAs associated with ischemia/reperfusion injury (IRI), as illustrated in the network. As is shown in the network, hsa-miR-135b-5p and hsa-miR-149-5p were the potential targets of both circFOXN2 and circNECTIN3. Subsequently, the target genes of the miRNAs were predicted by miRWalk and TargetScan. The circFOXN2/circNECTIN3-miRNA-

mRNA network is shown in Figure 5.

Discussion

EAD is a reversible injury following liver transplantation. Many factors contribute to the incidence of EAD, including graft IRI, hepatic steatosis, donor BMI, donor age and donor serum sodium level (6,18-20). Among these risk factors, graft IRI is generally considered one of the most important risk factors. Similar results have been found in the present study, showing that CIT was an independent risk factor for EAD. In addition to the clinical factors, Kurian *et al.* investigated the molecular mechanisms underlying EAD, defined relevant pathways (PPAR α

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Table 1 Univariate analysis of risk factors for EAD				
Variables	EAD, n=29	Non-EAD, n=86	P value	
Recipient age (years)	47.38±11.02	51.67±10.77	0.067	
Recipient sex			0.588	
Male	23 (79.31%)	72 (83.72%)		
Female	6 (20.69%)	14 (16.28%)		
Preoperative MELD score	37.00 (15.00)	21.00 (18.30)	0.000	
Operative time (h)	5.43±0.87	5.14±1.11	0.199	
Donor age (years)	46.00 (14.96)	42.96 (18.90)	0.203	
Donor sex			0.554	
Male	25 (86.21%)	70 (81.40%)		
Female	4 (13.79%)	16 (18.60%)		
Donor BMI (kg/m²)	22.88±2.79	22.59±2.75	0.627	
CIT (h)	12.38 (3.40)	7.21 (3.61)	0.000	
Macrovesicular steatosis			0.341	
<20%	26 (89.66%)	83 (96.51%)		
≥20%	3 (10.34%)	3 (3.49%)		
Two-circular RNA signature			0.000	
А	1 (3.45%)	31 (36.05%)		
В	7 (24.14%)	26 (30.23%)		
С	21 (72.41%)	29 (33.72%)		

Two-Circular RNA signature A: high circFOXN2 expression; B: low circFOXN2 expression and high circNECTIN3 expression; C: low circFOXN2 expression and low circNECTIN3 expression. EAD, early allograft dysfunction; MELD, Model for End-stage Liver Disease; CIT, cold ischemia time; BMI, body mass index.

Table 2 Multivariate analysis of risk factors for EAD

Risk factors	OR (95% CI)	Р	
CIT	1.498 (1.246–1.800)	0.000	
Two-circular RNA signature		0.013	
А	1.00		
В	22.439 (2.577–195.368)		
С	10.154 (1.041–99.006)		

Two-Circular RNA signature A: high circFOXN2 expression; B: low circFOXN2 expression and high circNECTIN3 expression; C: low circFOXN2 expression and low circNECTIN3 expression. CIT, cold ischemia time.



Figure 4 ROC curve analysis of the novel model combining twocircRNA signature indicates good predictive value for EAD (AUC: 0.870; 95% CI: 0.797-0.942).

and NF-KB) and targets (such as CXCL1, IL1, TRAF6, TIPARP, and TNFRSF1B) associated with the incidence of EAD (9). CircRNAs are often more stable than linear RNAs, which makes circRNAs promising biomarkers and therapeutic targets (12). However, the expression profile and predictive value of circRNAs in EAD remain unknown.

Previous research found that donor biology plays an important role in tumour recurrence and new-onset diabetes mellitus after liver transplantation (21,22). Therefore, donor biology is associated with the outcomes after liver transplantation. In this study, graft circRNA profiles were established through circRNA sequencing. We further identified that circFOXN2 and circNECTIN3 were significantly downregulated in the EAD group. Furthermore, we developed circular RNA signature (A, B and C) according to the expression of circFOXN2 and circNECTIN3. We also constructed a novel model combing circular RNA signature to predict EAD with good performance. The present study indicates that donor biology plays an important role in the development of EAD.

Previous studies have attempted to predict EAD. Hoyer et al. developed an EAD prediction score based on clinical variables of donors (6). A nomogram was also constructed based on the clinical parameters of donors and recipients to predict individual incidence of EAD (7). In addition to clinical variables, donor biology has been used to predict EAD. Kurian et al. defined a gene expression signature in donor liver tissue with high predictive value for EAD (9). Additionally, a recent study found that miR-146b-5p can act as a biomarker of EAD (8). However, the role of stable circRNAs, which are perfect candidate predictive indicators,

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Figure 5 CircFOXN2/circNECTIN3-miRNA-mRNA network construction. The red dots represent circFOXN2 and circNECTIN3; the green triangles represent the predicted miRNAs; and the blue quadrilaterals represent the target genes.

has not been elucidated before. The present study demonstrated the expression profile and predictive value of circRNAs in donor liver tissues.

In the present study, we selected miRNAs related to IRI and constructed a circFOXN2/circNECTIN3-miR-135b-5p/miR-149-5p network using database prediction. CircRNAs can act as miRNA sponges. MiRNAs are small, noncoding RNAs that play a pivotal role in the posttranscriptional regulation of gene expression. To explore the potential pathways related to EAD, we selected miRNAs associated with IRI to be construct a network. Many miRNAs, such as miR-21-3p and miR-490-3p, are involved in IRI (23,24). Regarding cardiac IRI, a recent study found that miR-149-5p could negatively regulate sphingosine-1-phosphate receptor 2 levels and improve the outcomes of transient middle cerebral artery occlusion in rats (25). Xie et al. also demonstrated that miR-135b-5p inhibition could protect cardiomyocytes from IRI in a mouse model through the JAK2/STAT3 pathway (26). Furthermore, Xing et al. found that miR-27a-5p was upregulated during hepatic IRI in mice (27). In addition, a previous study found that miR-133a-5p reversed the protective effect of propofol in a rat hepatic IRI model (28). Moreover, Li et al. illustrated that miR-370 levels were significantly increased in a mouse hepatic IRI model (29). Similar results have also been found in recent research (30). In the present study, ceRNA network analysis showed that circFOXN2 and circNECTIN3 regulated miR-135b-5p and miR-149-5p, which play important roles in IRI. Thus, the circFOXN2/

circNECTIN3-miR-135b-5p/miR-149-5p network provides a novel avenue for exploring the incidence of EAD.

Nevertheless, this study had several limitations. First, this is a single-centre retrospective study. Second, the sample size was relatively small. Further prospective multicentre studies with large samples are needed to validate the results.

In conclusion, this study demonstrated the differential expression of 442 circRNAs between the EAD and non-EAD groups. The novel model combing donor circular RNA signature is effective at predicting EAD. Additionally, the circFOXN2/circNECTIN3-miRNA-mRNA network provides potential targets for predicting and treating EAD.

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Footnote

Conflicts of Interest: The authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. This study was approved by the hospital ethics committee and complied with the Declaration of Helsinki. In addition, informed consent was obtained from all patients to allow the use of samples and clinical data for investigation. All collected samples were manipulated according to the protocols approved by the hospital ethics committee.

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Supplementary

Table S1 sequences of the primers

Primer	Sequence (5' to 3')	
GAPDH F	ATCCTGGGCTACACTGAGCA	
GAPDH R	AAGTGGTCGTTGAGGGCAAT	
CircFOXN2 F	CTGCACCAACAGGCTGGAAG	
CircFOXN2 R	TACTCTTACAGTTCTTGCCA	
CircNECTIN3 F	TCATCTACATTTCAGGTGCCT	
CircNECTIN3 R	TCTTCTCCCATGAAATCTGTG	