

Construction of a Necroptosis-Related miRNA Signature for Predicting the Prognosis of Patients With Hepatocellular Carcinoma

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Meng T, Wang Q, Yang Y, Ren Y and Shi Y (2022) Construction of a Necroptosis-Related miRNA Signature for Predicting the Prognosis of Patients With Hepatocellular Carcinoma. Front. Genet. 13:825261. doi: 10.3389/fgene.2022.825261 **Background:** Many miRNAs have been demonstrated to be associated with the prognosis of hepatocellular carcinoma (HCC). However, how to combine necroptosis-related miRNAs to achieve the best predictive effect in estimating HCC patient survival has not been explored.

Methods: The mRNA and miRNA expression profile were downloaded from a public database (TCGA-LIHC cohort). Necroptosis-related genes were obtained from previous references, and necroptosis-related miRNAs were identified using Pearson analysis. Subsequently, differential expression miRNAs (DEms) were identified in HCC and paracancer normal samples based on necroptosis-related miRNA expression. The whole set with HCC was randomized into a training set and testing set (1:1). LASSO-Cox regression analysis was used to construct an miRNA signature. Multiple statistical methods were used to validate the clinical benefit of signature in HCC patients, including receiver operator characteristic (ROC) curves, Kaplan–Meier survival analyses, and decision curve analysis (DCA). The downstream target genes of miRNAs were obtained from different online tools, and the potential pathways involved in miRNAs were explored. Finally, we conducted RT-qPCR in SK-HEP-1, THLE-3, and HUH-7 cell lines for miRNAs involved in the signature.

Results: The results showed that a total of eight specific necroptosis-related miRNAs were screened between HCC and adjacent tissues in the training set. Subsequently, based on the aforementioned miRNAs, 5-miRNA signature (miR-139-5p, hsa-miR-326, miR-10b-5p, miR-500a-3p, and miR-592) was generated by LASSO-Cox regression analysis. Multivariate Cox regression analysis showed that the risk scores were independent prognostic indicators in each set. The area under curves (AUCs) of 1 year, 3 years, 5 years, and 7 years were high in each set (AUC >0.7). DCA analysis also revealed that the risk score had a potential benefit than other clinical characteristics. Meanwhile, survival analysis showed that the high-risk group showed that specific miRNAs were mainly enriched in the cAMP signaling pathway and TNF signaling

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pathway. Finally, the results of RT-qPCR were consistent with the prediction results in public databases.

Conclusion: Our study establishes a robust tool based on 5-necroptosis-related miRNAs for the prognostic management of HCC patients.

Keywords: hepatocellular carcinoma, miRNAs, signature, necroptosis, prognosis

INTRODUCTION

By 2025, more than 1 million new cases of hepatocellular carcinoma (HCC) will be diagnosed each year worldwide, a serious situation that will pose a major challenge to global healthcare; more importantly, the 5-year survival rate for HCC patients has decreased by 20% globally and as low as 12% in Asian countries such as Japan. (Llovet et al., 2021). Smoking, drinking, and viral infection are risk factors for the occurrence and prognosis of HCC (El-Serag, 2011). In addition, local recurrence and metastasis reduce the survival rate in HCC patients (Di Sandro et al., 2019). Therefore, intricate etiological factors and heterogeneity of HCC make prognostic prediction challenging. There is an urgent need to develop a new prognostic model considering the limitations of HCC treatment strategies.

Necroptosis is a cell death independent of caspase (Linkermann and Green, 2014) including the following characteristics: incomplete cell membrane, intracellular metabolic abnormalities, and release of inflammatory factors. (Christofferson and Yuan, 2010). It is worth noting that necroptosis plays an important role in the occurrence and development of various diseases such as neurodegenerative diseases (Dionísio et al., 2020), ischemic cardiovascular (Ruan et al., 2019), and cancers (Gong et al., 2019). Interestingly, necroptosis has been shown to play a dual role in cancer. On the one hand, the hub regulators of necroptosis can promote metastasis and progression of cancer (Strilic et al., 2016); on the other hand, necroptosis can also prevent tumor development when apoptosis function in cancer cells is impaired (Long and Ryan, 2012). Meanwhile, necroptosis is regulated by intracellular signaling factors such as tumor necrosis receptor factor (TNFR) (Lalaoui et al., 2015), pattern recognition receptors (PRRs) (Li et al., 2012), and T-cell receptors (TCRs) (Hitomi et al., 2008). At present, relevant reports have revealed the regulation mechanism of necroptosis in HC. Xiang et al. (2021) discovered how CX32 induces necroptosis, and high CX32 expression may be represented as a resistant role to apoptosis inducers. Meanwhile, mosaic mouse models were used to reveal how necroptosis microenvironment directs lineage commitment in liver cancer (Seehawer et al., 2018). In addition, it is reported that some compounds can induce necroptosis to treat cancer. These evidence remind us that necroptosis has a potential application value in the treatment of HCC. In addition, microRNAs (miRNAs) are small molecules encoded by the genomes of eukaryotes, similar to siRNA (Bartel, 2004). There have been many studies on miRNA in HCC, such as the one where Wang et al. (2017a) found that miR-218 can suppress the metastasis and EMT of HCC cells via targeting SERBP1. Moreover, different

noncoding RNAs also interact with each other in the progression of HCC. Ding et al. (2020) revealed that hsa_circ_0001955 enhances proliferation, migration, and invasion of HCC cells through miR-145-5p/NRAS axis. Surprisingly, a combination of miRNA signatures in predicting HCC survival was elucidated by Li et al. (2020). In addition, enhancers can act as tissue-specific

| TABLE 1 | Clinicopathological | features of | TCGA-LIHC |
|---------|-----------------------|--------------|-----------|
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| Variables | Count | Percentage (% |
|--------------------|---------------|----------------|
| Age (mean ± SD) | 59.57 ± 13.33 | |
| Status | | |
| Alive | 249 | 66.05 |
| Dead | 128 | 33.95 |
| Gender | | |
| Male | 255 | 67.64 |
| Female | 122 | 32.36 |
| Pathological stage | | |
| Stage I | 175 | 46.42 |
| Stage II | 87 | 23.08 |
| Stage III | 3 | 0.80 |
| Stage IIIA | 65 | 17.24 |
| Stage IIIB | 9 | 2.39 |
| Stage IIIC | 9 | 2.39 |
| Stage IV | 2 | 0.53 |
| Stage IVA | 1 | 0.27 |
| Stage IVB | 2 | 0.53 |
| Unknown | 24 | 6.37 |
| T staging | | |
| T1 | 185 | 49.07 |
| T2 | 93 | 24.67 |
| T2a | 1 | 0.27 |
| T2b | 1 | 0.27 |
| T3 | 45 | 11.94 |
| T3a | 29 | 7.69 |
| T3b | 7 | 1.86 |
| T4 | 13 | 3.45 |
| TX | 1 | 0.27 |
| Unknown | 2 | 0.53 |
| N staging | 2 | 0.00 |
| NO | 257 | 68.17 |
| N1 | 4 | 1.06 |
| NX | 115 | 30.50 |
| Unknown | 1 | 0.27 |
| M staging | I | 0.27 |
| MO | 272 | 72.15 |
| M0 M1 | 4 | 1.06 |
| MX | 101 | 26.79 |
| Grade | 101 | 20.19 |
| GI2000 | 55 | 14.59 |
| G2 | 180 | 47.75 |
| G2 G3 | 180 | 47.75 32.89 |
| G3 G4 | 124 | 32.89 |
| G4 Unknown | 5 | |
| UNKNOWN | o | 1.33 |

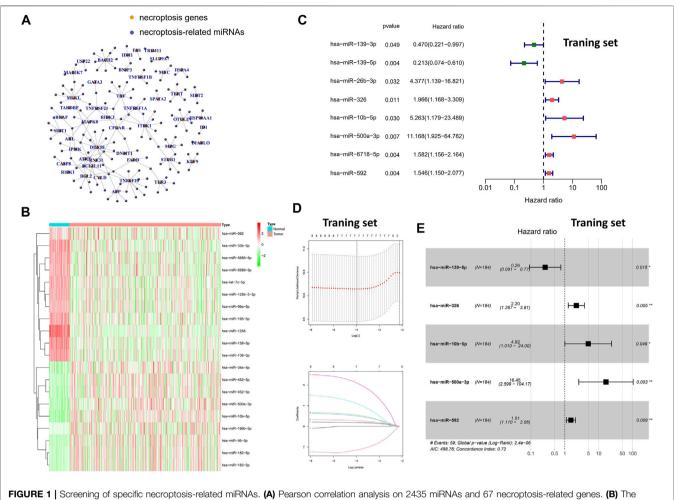


FIGURE 1 | Screening of specific necroptosis-related miRNAs. (A) Pearson correlation analysis on 2435 miRNAs and 67 necroptosis-related genes. (B) The heatmap of DEms in different tissues samples. (C) Univariate Cox regression analysis of 35-miRNAs (only the 8 miRNAs with statistical significance were shown). (D) LASSO regression analysis. (E) Multivariate Cox regression analysis in 8 prognostic miRNAs.

TABLE 2 | Results of multivariate Cox regression in 5-necroptosis-related miRNAs.

| miRNAs | Coef | HR | HR.95L | HR.95H | p |
|-----------------|---------|---------|--------|----------|--------|
| hsa-miR-139-5p | -1.3322 | 0.2639 | 0.0906 | 0.7687 | 0.0146 |
| hsa-miR-326 | 0.7874 | 2.1976 | 1.2674 | 3.8105 | 0.0051 |
| hsa-miR-10b-5p | 1.5938 | 4.9225 | 1.0097 | 23.9995 | 0.0486 |
| hsa-miR-500a-3p | 2.8005 | 16.4536 | 2.5987 | 104.1749 | 0.0029 |
| hsa-miR-592 | 0.4140 | 1.5128 | 1.1095 | 2.0627 | 0.0089 |
| | | | | | |

cis-regulatory elements to positively regulate gene expression by miRNAs (Zhang et al., 2020).

However, no study has systematically used necroptosis-related miRNA to predict the prognosis of HC patients.

In order to solve the aforementioned problem, we downloaded expression data and clinical features of TCGA-LIHC cohort from public database and extracted miRNA data related to necroptosis. Then, a 5-miRNA prognosis signature was constructed by LASOO-Cox regression analysis, and its prognostic ability was verified in different cohorts. The functional enrichment analysis of downstream genes of miRNAs is performed to explore the potential mechanism. Finally, we conducted RT-qPCR in SK-HEP-1, THLE-3, and HUH-7 cell lines for 5-miRNAs involved in the signature.

MATERIALS AND METHODS

Datasets and Pre-Processing

We downloaded the clinical features and miRNA expression of HC patients from The Cancer Genome Atlas (TCGA) database (Wang et al., 2016) *via* the R package "TCGAbiolinks". The miRNA expression profile in TCGA-LIHC includes 50 normal samples and 375 tumor samples (**Table 1**). Based on previous studies (Liu et al., 2021a; Visalli et al., 2018), we obtained 16 miRNAs associated with necroptosis, which were miR-495, miR-331-3p, miR-15a, miR-148a-3p, miR-75p, miR-141-3p, miR-425-5p, miR-200a-5p, miR-210, miR-223-3p, miR-500a-3p, miR-181-5p, miR-16-5p, hsa-miR-371-5p, hsa-miR-373, and hsa-miR-543 in detail. In addition, 67 necroptosis-related genes were extracted based on previous bioinformatics research (Hu et al., 2022; Huang et al.,

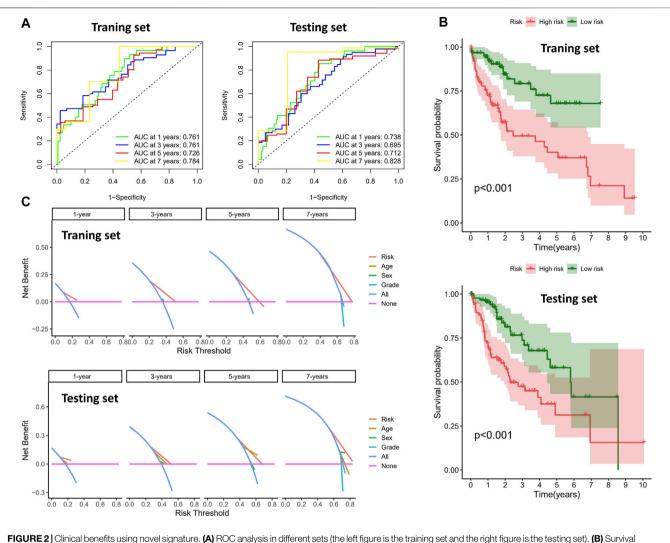


FIGURE 2 | Clinical benefits using novel signature. (A) ROC analysis in different sets (the left figure is the training set and the right figure is the testing set). (B) Survival analysis of different risk groups (the top figure is the training set and the bottom figure is the testing set). (C) DCA analysis in different sets (the top figure is the training set and the bottom figure is the testing set).

2021). We performed Pearson correlation analysis (|cor|>0.15, p < 0.05) on 2435 miRNAs and 67 necroptosis-related genes from the raw data [RNA-seq (log FPKM+1 format), miRNA-seq (log RPM +1 format)]. Ultimately, we annotated 144 necroptosis-related miRNAs in TCGA-LIHC from the references and Pearson correlation analysis.

Calculation of the Risk Score *via* the Necroptosis-Related MicroRNA Signature

The differentially expressed miRNAs (DEms) in normal samples and tumor samples were screened by using "limma" package (Ritchie et al., 2015) in R software. The thresholds in 'limma' package were set to adjusted *p*-value < 0.05, and |logFCfilter| = 1. At a ratio of 1:1, the whole set was divided into two using the 'caret' package in R software. We performed LASSO-Cox regression analysis in the training set (p < 0.05), and the risk score was calculated as follows:

$risk\ score = \sum_{i=1}^{n} Coef_i * x_i,$

where Coef is the coefficient, and x is the expression value of each selected miRNA. LASSO is a popular algorithm which was extensively utilized in medical studies (Liu et al., 2022a), (Liu et al., 2022b), (Liu et al., 2022c), (Liu et al., 2022d). The risk signature for predicting survival was assessed by the AUC value. We calculated the median score of risk score, which is used to select high-risk and low-risk groups in each sets. Kaplan–Meier survival analysis suggested the difference in high-risk and lowrisk groups. To better evaluate clinical applications, we calculated the net benefit rates of different clinical characteristics by the DCA curve analysis.

miRNA-mRNA Network

miRDB (Chen and Wang, 2020), TargetScan (Riffo-Campos et al., 2016), and MiTarBase (Huang et al., 2020) online tools were used to predict the downstream target genes of miRNAs

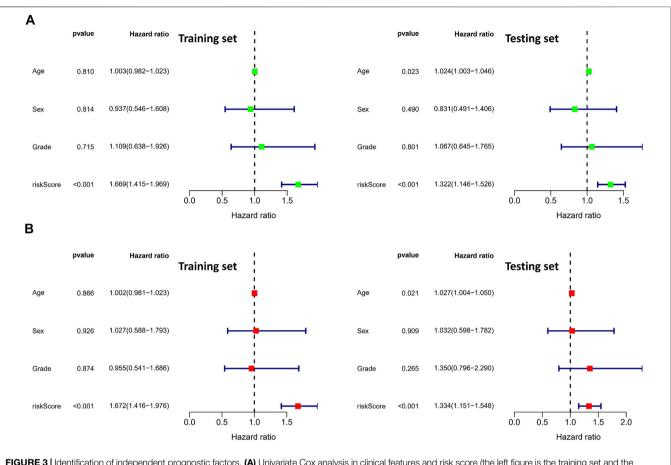


FIGURE 3 | Identification of independent prognostic factors. (A) Univariate Cox analysis in clinical features and risk score (the left figure is the training set and the right figure is the testing set). (B) Multivariate Cox analysis in clinical features and risk score (the left figure is the training set and the right figure is the testing set).

significantly related to prognosis, and only genes co-existing in three databases could be selected as the ultimate target genes. Finally, Cytoscape software was used to demonstrate the mRNA-target genes network.

Enrichment Analysis

GO enrichment analysis is a commonly used bioinformatics method, which is used to search for comprehensive information of large-scale genetic data. Meanwhile, the KEGG pathway enrichment analysis is widely used to understand biological mechanisms. The enrichment analysis was performed in downstream mRNAs by using "ggplot2" and "clusterProfiler" packages in R software (Yu et al., 2012).

Cell Culture and qRT-PCR

As seen in a previous study (Chen et al., 2020), HCC cell lines HUH-7 and SK-HEP-1 from humans and a normal human liver cell line THLE-3 were purchased from Shanghai Institute of Cell Biology and have been identified by STR genotyping test. These cells in a culture room (5% CO2 and 37°C) were cultured using RPMI-1640 medium with 10% FBS. We used miRcute miRNA Isolation Kit (Zhisheng, Nanjing, China) to isolate total miRNA. For miRNA, miRcute Plus miRNA First-Strand cDNA Synthesis kits (Zhisheng, Nanjing, China) were used for reverse transcription. The second step was completed using miRcute Plus miRNA qPCR Detection Kits (Zhisheng, Nanjing, China). Small RNA RNU6B (U6) (RiboBio, Guangzhou, China) was used as a control for the expression of miRNA (Li et al., 2022a). Primer sequences are summarized in previous studies, including miR-139-5p (Li et al., 2022b), hsa-miR-326 (Wei et al., 2022), miR-10b-5p (Niu et al., 2021), miR-500a-3p (Long et al., 2022), and miR-592 (Paul et al., 2021).

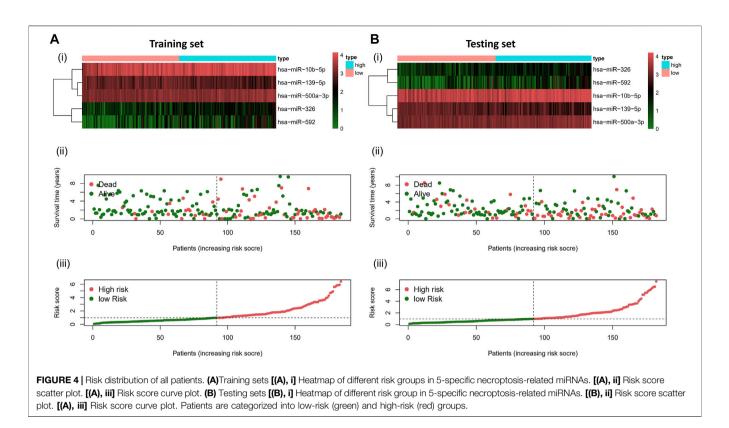
Statistical Analysis

All statistical analyses were performed using the R software (v.4.0.1). Detailed statistical methods about transcriptome data are covered in the bioinformatics method section. For the symbols, ***, **, *, and ns, refer to p < 0.001, <0.01, <0.05, and not significant, respectively.

RESULTS

Screening of Specific Necroptosis-Related MicroRNAs in Hepatocellular Carcinoma

Based on previous studies and Pearson correlation analysis, we obtained 144 miRNAs associated with necroptosis (Figure 1A). Subsequently, we performed differential expression analysis of



the aforementioned miRNAs in the TCGA-LIHC cohort by "limma" package. Finally, a total of 35 necroptosis-related miRNAs were identified in 50 normal samples and 375 tumor samples (Figure 1B). Compared with normal tissues, the aforementioned 35-miRNAs were abnormally expressed in HCC tissues, indicating that these miRNAs were worthy of further exploration. Subsequently, univariate cox regression was performed for further screening of 35-miRNAs (Figure 1C), and LASSO regression analysis further eliminated the redundant genes in 8 prognostic miRNAs (Figure 1D). Finally, multivariate COX regression analysis was performed for the aforementioned eight prognostic miRNAs, and the minimum Akaike information criterion (AIC) value was reached when miR-139-5p, hsa-miR-326, miR-10b-5p, miR-500a-3p, and miR-592 were included in the regression equation (Figure 1E).

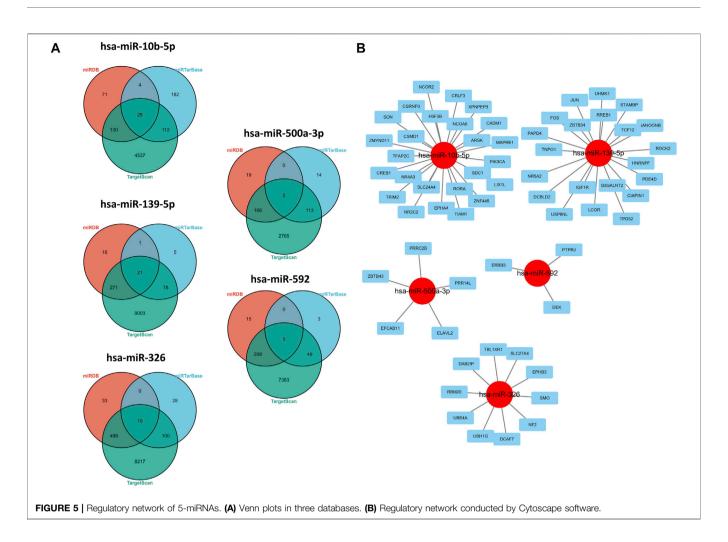
A Novel Signature Based on 5-Specific Necroptosis-Related miRNAs

We developed a novel signature based on the regression coefficients of 5-specific necroptosis-related miRNAs (**Table 2**), and all patients were divided into high-risk and low-risk groups according to the median value of the risk score in the training set. Hence, the formula of risk score = $(-1.3322 \times \text{expression level of hsa-miR-139-5p}) + (0.7874 \times \text{expression level of hsa-miR-326}) + (1.5938 \times \text{expression level of hsa-miR-10b-5p}) + (2.8005 \times \text{expression level of hsa-miR-500a-3p}) + (-0.4140 \times \text{expression level of hsa-miR-592})$. It is

worth mentioning that we also performed Kaplan–Meier analysis and log-rank test on 5-miRNAs in the whole set (**Supplementary Figure S1**). The results showed that low expression of the mir-139-5p group had less possibility of survival (p < 0.05). Taken together, our data showed that miR-139-5p, hsa-miR-326, miR-500a-3p, and miR-592 may have potential implications for the survival of HCC patients in the whole set.

Clinical Benefits of Using the Novel Signature in Patients With Hepatocellular Carcinoma

To validate the prognostic value of the risk score, we conducted survival analyses by plotting Kaplan-Meier curves and ROC analysis. As a result, the area under curves (AUCs) of 1 year, 3 years, 5 years, and 7 years were high in the training and testing sets (AUC = 0.761, 0.761, 0.726, and 0.784 in the training set; AUC = 0.738, 0.695, 0.712, and 0.828 in the testing set), as shown in Figure 2A. In addition, the survival rate can be separated between the high-risk and low-risk groups, and the high-risk group showed low survival probabilities (p < 0.001), as shown in Figure 2B. Interestingly, DCA analysis also revealed that the risk score had a net benefit value than other clinical characteristics (Figure 2C). To validate whether the risk score, patient age, tumor grade, and gender can function as independent predictors, we conducted univariate (Figure 3A) and multivariate Cox regression (Figure 3B) analyses for these clinical characteristics in the training set and testing set. The results showed that the risk score was an independent prognosis factor among the four clinical characteristics in multivariate analysis



(HR = 1.672 in training set, HR = 1.334 in testing set). The heatmap showed the expression of 5-specific necroptosis-related miRNAs in different risk groups [**Figures 4Ai, Bi**]. In addition, as the risk score increased, the patient death risk increased and the survival time decreased [**Figure 4Aii,iii, Bii,iii**].

Construction of Regulatory Network

As shown in the results of the aforementioned section, miR-139-5p, hsa-miR-326, miR-10b-5p, miR-500a-3p, and miR-592 may have potential implications for survival of HCC patients, so we explored the potential regulatory axis using miRDB, TargetScan, and MiTarBase online tools. Finally, 21, 10, 25, 5, and 3 potential downstream targeted genes were identified in miR-139-5p, hsamiR-326, miR-10b-5p, miR-500a-3p, and miR-592, respectively (Figure 5A). As shown in Figure 5B, we visualized the aforementioned network. More specifically, ARSK, CADM1, EPHA4, H3F3B, MAPRE1, NCOA6, NCOR2, NR2C2, NR4A3, PIK3CA, RORA, CREB1, CRLF3, CSMD1, CSRNP3, SLC24A4, SON, TFAP2C, TIAM1, TRIM2, XPNPEP3, ZMYND11, ZNF445, LIX1L, and SDC1 were the downstream targets of miR-10b-5p. ZBTB34, UHMK1, USP6NL, CIAPIN1, TPD52, IGF-1R, STAMBP, TCF12, DCBLD2, TNPO1, B3GALNT2, NANOGNB, NR5A2, PAPD4, PDE4D, ROCK2, RREB1, JUN,

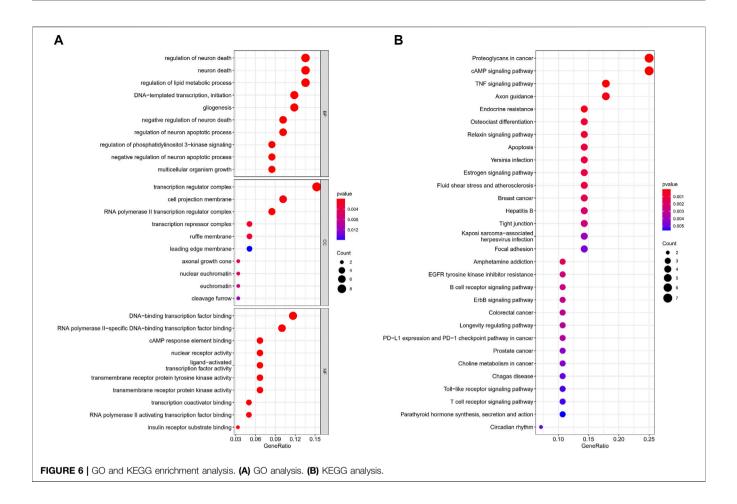
FOS, HNRNPF, and LCOR were the downstream targets of miR-139-5p. UBE4A, USH1G, DAB2IP, DCAF7, EPHB3, NF2, RBM20, SLC27A4, SMO, and TBL1XR1 were the downstream targets of miR-326. ZBTB43, PRR14L, PRRC2B, EFCAB11, and ELAVL2 were the downstream targets of miR-500a-3p. ERBB3, DEK, and PTPRJ were the downstream targets of miR-592.

Gene Enrichment Analysis

In order to further explore the biological functions involved in necroptosis-related miRNAs, we conducted an in-depth enrichment analysis. The results of GO enrichment analysis of the aforementioned targeted genes showed that they were mainly enriched in the regulation of neuron death, DNA-templated transcription, and gliogenesis. (**Figure 6A**). Meanwhile, KEGG enrichment further elucidated possible pathways such as proteoglycans in cancer, cAMP signaling pathway, and TNF signaling pathway. (**Figure 6B**).

qRT-PCR

The expression levels of the 5-miRNAs were determined using the TCGA database and qRT-PCR experiment. As shown in **Figure 7A**, the violin plot showed that tumor tissues had higher expression levels of miR-10b-5p and miR-500a-3p than normal

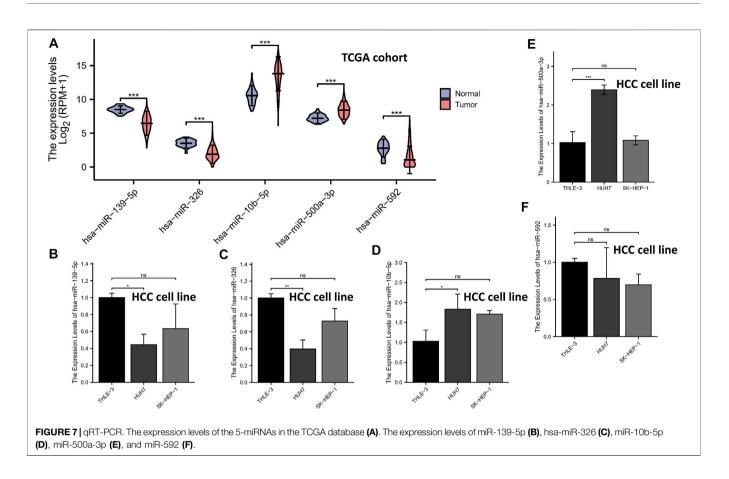


tissues, while normal samples had higher expression levels of miR-139-5p, hsa-miR-326, and miR-592 compared to HCC samples. Subsequently, HCC cell lines (HUH-7 and SK-HEP-1) and a normal human liver cell line (THLE-3) were used to validate the results of tissue samples. In the validation of miR-139-5p (Figure 7B), hsa-miR-326 (Figure 7C), miR-10b-5p (Figure 7D), and miR-500a-3p (Figure 7E), the expression of HUH-7 was higher than that of THLE-3. However, there was no statistical difference in the expression of the aforementioned miRNAs in SK-HEP-1 vs. THLE-3. In the validation of miR-592, no significant differences in the expression of all three cell lines were observed (Figure 7F). Overall, most miRNA expressions were at the same level in cell lines and tissues, except for miR-592.

DISCUSSION

Although there are many studies on miRNA expression to predict the prognosis of patients, no study has systematically used necroptosis-related miRNA to predict the prognosis of HCC patients. To our knowledge, our study is the first to explore the clinical application of necroptosis-related miRNAs in an HCC cohort. In this study, a total of 5-miRNA signatures were generated by Cox regression. Subsequently, a series of statistical analyses demonstrated the excellent clinical application for the risk score calculated based on the risk signature. Overall, most miRNA expressions were at the same level in cell lines and tissues, except for miR-592. Moreover, we found that specific miRNAs were mainly enriched in the cAMP signaling pathway, TNF signaling pathway, and Wnt/β-catenin pathway. Fortunately, the role of the aforementioned pathways in the pathogenesis of HCC has been verified in in vitro experiments. Liu et al. (2021b) revealed that cellular retinol binding protein-1 inhibits cancer stemness via upregulating WIF1 to suppress the Wnt/β-catenin pathway in hepatocellular carcinoma. Meanwhile, Yassin et al. (2022) explored silvbum marianum total extract, silvmarin and silibinin abate hepatocarcinogenesis, and hepatocellular carcinoma growth via modulation of the TNF signaling pathway. In addition, activation of the cAMP signaling pathway in HCC also appeared to be associated with epigenetic modifications, as in Cai et al.'s (2021) study that RBM15-mediated m6A modification might facilitate the progression of HCC via the IGF2BP1-YES1-MAPK axis. This seems to suggest that there is cross-talk in the pathways that promote the development of HCC. Thus, this study lays the foundation for our future in vivo and in vitro experiments.

Most of the miRNAs participating in signatures have been revealed to be associated with cancer progression. In miR-139-5p, it suppressed the proliferation and migration of hepatocellular carcinoma cells by downregulating the expression of ENAH



(Zhang et al., 2022). Meanwhile, miR-326 can suppresses the Hippo pathway when combined with PAX8 in trophoblast (Zang et al., 2022). Moreover, Wang and Chen (2022) also revealed that overexpression of miR-362 inhibited the expression of JMJD2A in nasopharyngeal carcinoma, and aberrant miR-362 may be associated with EBV-infection. In miR-10b-5p, it can inhibit tumorigenesis in the gastric cancer xenograft mice model by downregulating Tiam1 (Liu et al., 2021c); therefore, Yan et al. (2021) put forward a novel idea for exosomal miR-10b-5p in gastric cancer. A novel conclusion was that upregulated exosomal miR-10b-5p is involved in fibroblasts in tumor microenvironment. In miR-500a-3p, it has been shown that silence miR-500a-3p may serve as a new therapeutic strategy in the treatment of HCC (Jiang et al., 2017). Currently, there are three studies on the mechanism of miR-592 in HCC, in which the downstream consists of IGF-1R (Wang et al., 2017b), WSB1 (Jia et al., 2016), and DEK (Li et al., 2015). In addition to the miRNAs involved in the model, other miRNAs have also been studied in depth in HCC. In some studies about miR-223-3p, it can be used as a novel noninvasive biomarker for HCV-positive cirrhosis and HCC (Oksuz et al., 2015). Circulating mirR-223-3p can represent novel diagnostic and prognostic markers for HBV-associated HCC patients (Pratedrat et al., 2020). It is interesting to note that miR-331-3p (AUC: 0.832) has better diagnostic performance than AFP (Jin et al., 2019). On the other hand, these evidence suggest that sequencing of blood samples from HCC patients may play a noninvasive role in predicting survival outcomes. The novel

mechanism of miR-500a-3p promoting HCC stem cell maintenance suggests that miR-500a-3p may be a novel therapeutic strategy (Jiang et al., 2017). Unfortunately, the critical role of the aforementioned pathways in HCC necroptosis is not well explained at present.

However, this study only used the data from the public database TCGA to construct the model, and there was no condition to collect our data to validate 5-miRNA signature, which was a limitation to our study. In addition, we have no conditions to verify the mechanism of these necroptosis-related miRNAs. In the future, we can use risk signatures based on altered 5-miRNAs, which may improve prognostic prediction and may lead to development of targeted therapy.

CONCLUSION

Comprehensively, our study suggests that necroptosis-related miRNAs is closely associated with the prognosis of HCC, and we established a robust tool for the prognostic management of HCC patients.

DATA AVAILABILITY STATEMENT

Publicly available datasets were analyzed in this study. These data can be found at: https://portal.gdc.cancer.gov/.

ETHICS 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.

AUTHOR CONTRIBUTIONS

TM and QW have contributed equally to this work and share first authorship. TM, QW, YY and YS conceived, designed, and supervised the study. TM and YR performed data analysis and drafted the manuscript. YS and YR helped revise the manuscript. TM, QW and YY collected the data. All the authors reviewed and approved the final manuscript.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fgene.2022.825261/full#supplementary-material

Supplementary Figure S1 | Kaplan-Meier analysis about 5-specific necroptosisrelated miRNAs in the whole set. miR-10b-5p (A), miR-139-5p (B), hsa-miR-326 (C), miR-500a-3p (D), and miR-592 (E).

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