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ORIGINAL RESEARCH

Transcriptome Profiles Characteristics of the Peripheral Immune in Patients with Severe Fever with Thrombocytopenia Syndrome

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Purpose: Severe fever with thrombocytopenia syndrome (SFTS) is an acute viral infection disease with a high mortality, but there are no specific effective drugs or vaccines available for use. To develop effective treatment methods, more basic researches are urgently needed to elucidate the response mechanisms of patients.

Patients and Methods: Here, we conducted the transcriptomic analysis of peripheral immunity in 14 SFTS patients, ranging from moderate infection to severe and fatal disease.

Results: The results showed orderly cytokine signaling pathway modulation in moderate patients, cellular immunosuppression in severe patients, and significant dysregulation of the inflammatory response and coagulation dysfunction characteristic of deceased patients. In addition, WGCNA further showed a significant positive correlation between fatal outcomes and B cell and immunoglobulin mediated immune function modules, as well as a significant negative correlation with coagulation function modules.

Conclusion: Overall, our research findings systematically observed potential immune mechanisms underlying clinical symptom heterogeneity and noteworthily revealed multiple signaling pathways leading to coagulation dysfunction in fatal outcomes, not just related to decreased platelet count, which can further elucidate the interaction between viruses and hosts and contribute to clinical treatment.

Keywords: SFTS, heterogeneous, fatal case, RNA-seq, peripheral immunity

Introduction

Severe fever with thrombocytopenia syndrome (SFTS), a new hemorrhagic fever first reported in China in 2009,¹ is an acute infectious disease caused by infection with the Dabie bandavirus (formerly known as severe fever with thrombocytopenia syndrome virus [SFTSV]), with a mortality rate that can be as high as 12–50%.^{2,3} The main transmission vectors of SFTSV are ticks, especially H. longicornis,⁴ and the blood and secretions of the patients can also cause humanto-human transmission and even trigger cluster outbreaks.^{5,6} In 2018, the World Health Organization listed SFTS as one of the diseases that pose major public health risks. In recent years, the geographic distribution of SFTS has been expanding, and the incidence of SFTSV has been reported in several Asian countries,^{7–9} but there is still no approved vaccine or specific antiviral treatment. The clinical symptoms of SFTS mainly include influenza-like symptoms such as acute fever, weakness, and headache, as well as gastrointestinal symptoms such as anorexia and diarrhea, accompanied by typical symptoms of leukopenia and thrombocytopenia. As the disease progresses, some patients may experience shock, hemorrhage, disseminated intravascular coagulation (DIC), and even multiple organ failure (MOF).¹⁰ According to the patient's symptoms and the severity of organ damage, it can be categorized into mild, moderate, severe, and critical types. Among them, patients with mild and moderate types usually have a platelet drop (higher than 75×10^9 /L) and have a good prognosis. On the other hand, the severe and critical types show progressive platelet decline (less than 50×10^9 /L or 25×10^9 /L), multiple organ damage, and are very likely to have a poor prognosis. However, the underlying mechanisms of the different clinical manifestations and treatment outcomes are largely unknown.

In general, the ability of viruses to survive, replicate, and spread in the host is not only dependent on the adaptive changes caused by virus evolution but also on the response and alteration of the host immune system and other complex internal environments. Therefore, exploring the molecular interactions that occur between SFTSV and the host and tapping into the mechanisms by which SFTSV evades the host immune response may help in the development of vaccines or therapeutics against viral infection. Previous studies have shown that some degree of immune dysfunction, such as monocyte abnormalities,¹¹ cytokine storms,¹² cellular immune dysregulation,¹³ and humoral immune deficiency,¹⁴ exists in severe and critical types or deceased patients. These observations have increased our understanding of the pathogenic of SFTS, but the exact mechanisms underlying the differential clinical manifestations remain elusive, and the transcriptional landscapes and gene expression profiles of SFTS patients in different disease states are not yet completely clear.

In recent years, transcriptomics has been widely applied to identify the immune response dynamics and gene regulatory networks mediated by viral infection.^{15–18} In this study, we recruited SFTS patients and grouped them according to their clinical symptoms and prognostic outcomes, systematically scanned the transcriptomic profiles of the patients, and parsed the underlying mechanisms of clinical heterogeneity, which provided a theoretical basis for clinical treatment and regression warning.

Materials and Methods

Study Population and Sample Isolation

From July to September 2023, we invited patients hospitalized with SFTSV infection confirmed by RT-PCR results and healthy individuals in Jinan and Yantai City of Shandong Province. The inclusion criteria for SFTS patients included two requirements: (1) patients with SFTSV infection confirmed by RT-PCR results; (2) The patient has no other immune related underlying diseases. The inclusion criteria for healthy individuals are gender and age comparable to the patient, had no immune-related underlying disease, and were not receiving immunotherapy. The anticoagulant venous blood samples were collected from healthy individuals and the patients within 7 days of onset, and peripheral blood mononuclear cells (PBMCs) were isolated by density-gradient sedimentation for further study.

Transcriptome Sequencing

The high-throughput transcriptome sequencing of participants was performed according to our previous research methods.¹⁹ The RNeasy Mini Kit (Qiagen, Germany) and Stranded Total RNA Prep Ligation with Ribo-Zero Plus Kit (Illumina, USA) were used to extract total RNA from PBMCs and prepare the rRNA-depleted cDNA library. The sequencing was carried out to generate 150 bp paired-end reads on the Illumina NextSeq 2000 platform. The quality control, trimming, and mapping of the RNA-seq raw FASTQ data to the human reference genome hg38 were performed using CLC Genomics Workbench (version 21.0). The gene expression level was measured based on the transcripts per million (TPM).

Differentially Expressed Genes (DEGs) Identification

To identify transcriptional features in peripheral immunity that correlate with clinical features of SFTS patients, we compared DEGs in PBMC between the patient groups and the healthy group. The normalization factors were calculated using edgeR²⁰ and limma²¹ packages, and the standardization and filtering of gene expression were accomplished by the

voom and eBayes functions. The p-value < 0.05 and $2^{\wedge logFC_cutoff}$ criteria were used to filter out DEGs. The fold change and expression profile of DEGs were visualized as volcanoes and heatmaps in R (version 4.1.0). The transcriptional profile of cytokine-related genes were also visualized by heatmap.

Prioritization of DEGs by Protein–Protein Interaction (PPI) Network Analysis

The initial PPI networks for the protein products of up-regulated and down-regulated DEGs were constructed from the STRING database (version 11.5),²² and visualized by Cytoscape software (version 3.9.1).²³ The MCODE plugin in Cytoscape software was used to identify the characteristic DEGs from the initial PPI networks.





Functional and Pathway Enrichment Analysis

The prospective gene ontology (GO) terms of DEGs from different groups were identified using clusterProfiler packages (v.4.8.3).²⁴ The Bonferroni-adjusted p-value <0.05 was used as the cut-off criterion.

Weighted Gene Co-Expression Network Analysis (WGCNA)

The weighted gene co-expression network analysis based on transcriptional profiles and sample characteristics was executed through the WGCNA package.²⁵ The specific details were consistent with our previous research. First of all, the



Figure 2 Volcano plot of global gene expression changes in the moderate (A), severe (B), and fatal (C) groups relative to healthy control. Red: up-regulated DEGs in the patients group compared with the healthy group; Blue: down-regulated DEGs in the patients group compared with the healthy group; Grey: non-DEGs. Criteria: |Fold Change|= 2^{NogFC} _cutoff is marked out by two black dotted lines. P-value = 0.05 is marked out by a horizontal black line y = $-Log 10^{(0.05)}$.

best β value was confirmed with a scale-free fit index larger than 0.80 as well as the highest mean connectivity. Subsequently, the topological overlap matrix (TOM) transformed by the adjacency matrix was then clustered by dissimilarity between genes, and then we performed hierarchical clustering to identify modules, each containing at least 30 genes (minModuleSize = 30). Finally, the co-expressed genes were determined by calculating the module membership (MM) and gene significance (GS) of the genes in the target modules. The functional enrichment of remarkable modules was carried out by the ClueGO plugin²⁶ in Cytoscape software (version 3.9.1). The Bonferroni-adjusted p-value <0.05 was also used as the cut-off criterion.

Results

Characteristics of Participants

To better understand the underlying mechanisms of clinical heterogeneity in SFTS patients, we performed transcriptome sequencing and analysis of PBMCs from 14 patients with SFTS confirmed by RT-PCR and 6 age-matched healthy controls. Based on clinical manifestations, all patients were further divided into moderate (n=5), severe (n=6), and fatal (n = 3) groups. The platelet counts of patients in the moderate group were all higher than 75×10^9 /L, while those in the



Figure 3 Continued.



Figure 3 Hierarchical cluster analysis of DEGs. The expression profile of the top 50 up-regulated and down-regulated DEGs in moderate (A), severe (B), and fatal (C) groups relative to healthy control. Each row represents mRNA and each column represents a sample. Red indicates higher expression and blue indicates low expression in the SFTS patients group as compared with that in the healthy group.

severe group were all lower than 50×10^9 /L. The gender and age characteristics of all participants and study design are shown in Figure 1.

Transcriptional Profile of SFTS Patients

We summarized the DEGs of SFTS patients with different clinical symptoms relative to healthy control. The volcano plot showed that 459 genes were up-regulated (eg, *IF127, IGLL5, BOLA2, TNFRSF17*, and *MZB1*) and 154 genes were down-regulated (eg, *ADRA2B, FCER1A, PF4V1, CXCL5*, and *RASD1*) in the moderate patients (Figure 2A). In total, 717 DEGs were identified in the severe group, including 464 up-regulated and 253 down-regulated DEGs (Figure 2B), and 447 up-regulated and 332 down-regulated DEGs were identified in the fatal group compared with the uninfected individuals (Figure 2C). The expression profile of the top 50 up-regulated and down-regulated DEGs of each sample were shown by heatmap (Figure 3A–C).

Orderly Antiviral Responses in the Moderate Group

To analyze the interaction relationships between up-regulated and down-regulated DEGs identified from the moderate group, the characteristic gene subnetworks contained 95 and 52 DEGs, respectively (Figure 4A and B). In order to clarify

Α	MARCO	CRISP3	LGALS3BP	IFI27L1	CHI3L1	CD70	DEFA3	CCL25	POU2AF1
	FPR2	AIM2	IGLL5	RNASE2	SLPI	OLFM4	CCR10	MZB1	DEFA1B
	S100A8	RNASE3	MERTK	CD177	TIMD4	AXL	C1QC	DEFA4	IGF1
	CXCR6	CCL7	SPATS2L	JCHAIN	XBP1	TNFRSF17	PGLYRP1	PRTN3	HP
	LAG3	C1QB	BATF2	S <mark>100</mark> A9	PDCD1LG2	IL1RN	LAP3	CXCR3	RETN
	CEACAM8	C1QA	TREM2	P <mark>PAR</mark> G	MMP8	H <mark>ERC</mark> 6	C <mark>MPK</mark> 2	IFITM3	ARG1
	PLSCR1	CAMP	ELANE	CTLA4	SERPING1	HERC5	EPSTI1	CD27	CSF1
	XAF1	USP18	IFI27	LY6E	IFITM1	IFI6	CCL8	LTF	IFI44L
	MPO	CD38	OAS3	IFI35	CCR5	OAS2	OASL	IFIT1	IF144
	GBP1	IFIT3	IFIH1	RSAD2	FCGR1A	MX1	SIGLEC1	ISG15	ММР9
	CXCL11	FCGR3B	IL10	CCL2	CXCL10				



Figure 4 Continued.



Figure 4 The PPI network extracted from initial PPI networks for the protein products of up and down-regulated DEGs induced by the moderate group, consisting of 95 nodes (\mathbf{A}) and 52 nodes (\mathbf{B}). The color and size of points demonstrate hubness of the node in the PPI network. PPI, protein–protein interaction; DEGs, differentially expressed genes. (\mathbf{C}) The dot plot of the top 15 up-regulated enriched terms in the biological process identified in the moderate group. (\mathbf{D}) The bar plot of the top 15 down-regulated BP-enriched terms identified in the moderate group. Criteria: p-value < 0.05.

the functional pathways of the characteristic DEGs, the top 15 GO-terms enriched by SFTSV were shown in the GO dot plot and bar plot. The up-regulated DEGs performed some virus defense functions, including "cytokine-mediated signaling pathway", "defense response to virus", "humoral immune response", "negative regulation of virtual process", and "negative regulation of virus genome replication", which are classic antiviral pathways (Figure 4C). The downregulated DEGs are involved in numerous immunomodulatory pathways, such as "chemokine-mediated signaling pathway", "response to chemokine", "neutrophil migration". The inhibition of cytokine-related signaling pathways may help avoid excessive inflammatory response and reduce the risk of inducing cytokine storms (Figure 4D). The reasonable combination of these activated and inhibited immune functional pathways may be a potential mechanism for the mildness of symptoms in some SFTS patients.

SFTSV Triggers Cellular Immune Suppression in the Severe Group

A similar characteristic DEG extraction was performed on transcriptome data in the severe group, 102 up-regulated and 67 down-regulated DEGs were retained, respectively (Figure 5A and B). Gene ontology (GO) analysis showed that upregulated DEGs in the severe group were mainly involved in the "defense response to virus", "negative regulation of viral genome replication", "humoral immune response", "cytokine-mediated signaling pathway", and "negative regulation of viral process", playing a positive role in the antiviral process (Figure 5C). It is noteworthy that down-regulated DEGs are not only in the "chemokine mediated signaling pathway" but also in the "regulation of CD4-positive, alphabeta T cell activation and differentiation" (Figure 5D). The suppression of cellular immune function is likely to be one of the immune escape mechanisms for the virus, which in turn leads to exacerbation of clinical symptoms in patients.

Α									
C2	GPRC5D	FOLR3	CCL25	MARCO	CCR10	MS4A4A	CTSL	DEFB1	FPR3
IL1R2	CFB	MERTK	IFI27L1	AIM2	ZBP1	DEFA1B	TCN1	MS4A3	RNASE2
FCRL5	POU2AF1	CEACAM6	RNASE3	S100A8	OLFM4	AXL	SLPI	DEFA1	TNFRSF13B
FPR2	IGLL5	CD177	CXCR3	MZB1	IGF1	SP <mark>ATS</mark> 2L	DEFA4	TNFRSF17	S100A9
PGLYRP1	XBP1	MUC1	LOC102723407	LAP3	HP	PRTN3	JCHAIN	C1QB	IFIT1B
BATF2	CEACAM8	C1QA	I <mark>SG2</mark> 0	IRF4	SDC1	MMP8	CD27	PPARG	CCR2
CMPK2	PLSCR1	PARP9	ELANE	IFITM3	SERPING1	EPSTI1	CCR5	MX2	XAF1
IFITM1	CD38	RTP4	HERC5	МРО	IFI6	IFI44L	IFI27	LTF	USP18
IFI35	OAS2	IFIT3	ММР9	FCGR1A	OASL	OAS3	GBP1	OAS1	CD163
IFIT1	IFIH1	IFI44	CXCL11	SIGLEC1	RSAD2	MX1	FCGR3B	ISG15	FN1
IL10	CCL2								

В

ARC	ERRFI1	F13A1	TNFRSF25	GADD45G	KLRC3-2	NR4A3	RGS2
IGFBP3	FGFBP2	TUBB1	MPL	TNFSF4	IER3	IER2	SMAD7
CCL3L1	FCER1A	DDIT3	SH2D1B	PPP1R15A	SOCS1	DUSP2	CLEC4C
PTGDR2	NFKBIZ	PF4V1	CXCL5	KIR3DL1	CD1E	FGF18	SNAI1
РРВР	CD83	TCF7	PF4	KLRF1	CD160	NR4A2	B3GAT1
CD244	GADD45B	JUNB	NCR3	KLRG1	CXCL2	NCR1	ATF3
ІТСВЗ	CD1C	FOSB	NR4A1	KLRK1	ERBB2	CCL3	PTGS2
CXCR5	KLRB1	NCAM1	CD69	CCL5	EGR1	CXCL8	JUN
IL1B	IFNG	TNF					

Figure 5 Continued.



Figure 5 The PPI network extracted from initial PPI networks for the protein products of up and down-regulated DEGs induced by the severe group, consisting of 102 nodes (A) and 67 nodes (B). The color and size of points demonstrate hubness of the node in the PPI network. PPI, protein–protein interaction; DEGs, differentially expressed genes. (C) The connection of the top 15 up-regulated enriched terms in the biological process with specific DEGs identified in the severe group. (D) The dot plot of the top 15 down-regulated BP-enriched terms in the severe group. Criteria: p-value < 0.05.

Absent Immune Regulation and Impaired Coagulation Function in the Fatal Group

To investigate the possible biological interactions of DEGs and identify important functional networks, we identified 80 up-regulated and 76 down-regulated characteristic DEGs in the fatal group (Figure 6A and B). Peripheral immunity showed that the most strongly regulated biological groups are related to extensive induction of antiviral signaling pathways, activation of the cytokine-mediated signaling pathway, and humoral immune response (Figure 6C). But what is completely different from the moderate and severe groups is that down-regulated DEGs are no longer widely involved in immune regulation, especially in the inhibition of cytokine responses, but are strongly enriched in "blood coagulation", "homeostasis", "platelet activation", and "regulation of cell-cell adhesion" (Figure 6D). This likely explains the dysregulated excessive inflammatory response in SFTS fatality. It is also emphasized that the hemorrhagic manifestations of the patients may be importantly linked not only to the apparent decrease in platelet counts but also to the underlying dysfunctions of coagulation and hemostasis.

The Biological Functions of Fatal Outcome-Related Gene Modules

The most suitable β value ($\beta = 10$) was determined as a soft threshold parameter to construct a scale-free network using the WGCNA package (Figure 7A and B). In the transcriptome data, 31 co-expression gene modules were identified by hierarchical clustering, and the dendrogram was clustered based on the dissimilarity measure (1-TOM) (Figure 7C and D). These modules based on differences in topological overlap were displayed in the 3D clustering map (Figure 7E and F). The correlations between the sample characteristics and the co-expression module were shown in Figure 7G, where the yellow module (eigengene value = 0.78) was significantly positively correlated with fatal SFTS. The brown module (eigengene value = -0.83) was significantly negatively correlated with the severity of SFTS. The immunity enrichment analysis showed that the yellow module was mainly involved in "positive regulation of B cell mediated immunity", "positive regulation of immunoglobulin mediated immune response", and "immunoglobulin production involved in immunoglobulin-mediated immune response" (Figure 8A and C). The significantly negatively correlated brown module tended to execute functions, including "response to interferon-alpha", "interleukin-27-mediated signaling pathway", "response to virus", and "platelet activation" (Figure 8B and D).





Figure 6 Continued.



Figure 6 The PPI network extracted from initial PPI networks for the protein products of up and down-regulated DEGs induced by the fatal group, consisting of 80 nodes (A) and 76 nodes (B). The color and size of points demonstrate hubness of the node in the PPI network. PPI, protein–protein interaction; DEGs, differentially expressed genes. (C) The bar plot of the top 15 up-regulated BP-enriched terms in the fatal group. (D) The connection of the top 15 down-regulated enriched terms in the biological process with specific DEGs identified in the fatal group. Criteria: p-value < 0.05.

The Cytokines Transcriptional Profile of SFTS Patients

Given the lethal role of cytokine storm in SFTS patients, we have mapped the transcription profiles of 67 cytokine-related genes. The clustering analysis results indicate that 23 cytokine-related genes, including *IL6, IL10, IL13, IL18,* and *IL19,* are highly expressed in the SFTS patient group, especially in the fatal group. Forty-four cytokine-related genes, including *IL1A, IL1B, IL2, IL4, CXCL8* (encoding IL-8), *TGFB1* (encoding TGF- β), etc., showed low expression in the fatal group (Figure 9).

Discussion

With the expansion of geographical distribution, SFTS with high lethality poses an imminent threat to human and public health, but our understanding of its clinical heterogeneity remains limited. Studying the transcriptional response of patients to the virus and discussing the impact of host regulation on clinical manifestations and outcomes is crucial for unraveling viral pathogenesis. In this study, the integrating bioinformatics method was applied to assess the transcriptional response profiles of SFTS patients and portray the transcriptional characteristics of the patients. Our data reveal the complex molecular response mechanisms of the host, and in particular summarize key molecular features associated with lethal outcomes, providing a valid basis for the development of therapeutic approaches against this viral infection.

The cytokine-mediated inflammatory responses play crucial roles in disease progression during SFTSV infection.^{27,28} Appropriate cytokines can enhance immune responses, while dysregulated cytokines can lead to an imbalance between pro-inflammatory and anti–inflammatory effects, causing immune damage.¹⁰ Recent theoretical developments have revealed that the severity of SFTS is associated with uncontrolled inflammatory response,²⁹ and have identified that cytokine storms are the main pathological feature of fatal SFTS patients.³⁰ However, the molecular mechanisms underlying the relationship between cytokines or chemokines and disease severity in SFTSV infection have not been well resolved. We found that the up-regulated and down-regulated DEGs in both moderate and severe patients jointly regulate the cytokine-mediated signaling pathway, which may play a key role in maintaining a balance between pro-inflammatory and anti-inflammatory effects and contributing to patient recovery. In contrast, the significant absence of anti-inflammatory effects in deceased patients may be an important factor leading to the hyperinflammatory features and pathologic damage. In addition, consistent with the previous two studies, there was significant overexpression of *IL6* and *IL10* and low expression of *TGFB1* in the fatal group, which is likely associated with cytokine storm and lethal

outcomes.^{31,32} These results suggest that appropriate immunosuppressive therapy may help to improve inflammation and pathological damage in SFTS patients and deserve further investigation.

There are growing appeals for the crucial role of T cell dysregulation in SFTS. Specifically, patients in the poor prognosis group have lower numbers of CD3⁺, CD3⁺ CD4⁺, and CD3⁺ CD8⁺ T cell clusters than healthy controls.³³ Compared with those in surviving patients, CD4⁺ T-cell depletion, Th1, Th2, and Treg numbers decreased were observed in deceased patients.³⁴ The above research results reveal the cellular immune dysregulation in SFTS patients from the perspective of cell subpopulation counting. In our study, we further discovered the inhibition of T cell activation and



Figure 7 Continued.



Figure 7 WGCNA of the PBMC transcriptome. (**A** and **B**) Analysis of network topology for various soft-thresholding powers. The left panel shows the scale-free fit index (y-axis) as a function of the soft-thresholding power (x-axis). The right panel displays the mean connectivity (degree, y-axis) as a function of the soft-thresholding power (x-axis). (**C**) Clustering dendrogram of genes, with dissimilarity based on topological overlap, together with assigned module colors. (**D**) The heatmap depicts the Topological Overlap Matrix (TOM) of genes selected for weighted co-expression network analysis. Light color represents lower overlap and red represents higher overlap. (**E** and **F**) 3D cluster map of genes, based on differences in topological overlap, and assigned module colors. (**G**) Module-trait associations: Each row corresponds to a module eigengene and each column to a trait. Each cell contains the corresponding correlation and p-value.

differentiation function in severe patients, which are mainly executed by *SMAD7*, *CD83*, *JUNB*, *SOCS1*, *IFNG*, *TNFSF4*, *NFKBIZ*, and *CD160*, revealing the potential mechanism of cellular immune damage from a molecular perspective.

In our study, all patients showed activation of humoral immunity compared to healthy controls. Strikingly, WGCNA showed that gene modules significantly positively associated with lethal outcomes were also extensively involved in B cell and immunoglobulin mediated immune response. Before this, several studies have also explored the role of humoral immunity in SFTS. For example, the percentage of B cells in the poor prognosis group was significantly higher than that in the recovery group,³³ and the proliferation of peripheral plasmablasts in lethal SFTS was higher than that in self-



Figure 8 (A and B) Scatter diagram for MM vs GS in the yellow and brown module. (C) The immunity functional enrichment analysis of the yellow gene module. (D) The GO functional enrichment analysis of the brown gene module. Criteria: p-value < 0.05.

limiting SFTS but showed dysfunctional features.³⁵ Autopsies following fatal human studies have identified B-cell lineage plasmablasts as one of the major target cells for SFTSV infection in fatal SFTS,³⁶ and a scRNA-seq study also found that IFN-I-inducible plasmablast expansion characteristics are associated with fatal outcome in SFTS.³⁷ Thus, it can be seen that adaptive B cell immunity is of great significance for the successful clearance of viruses, but dysregulated humoral immunity and the ensuing immunopathologic impairment are likely to lead to fatal outcomes. The monitoring and functional regulation of B cell clusters were expected to become key signals for predicting patient severity and important means of improving treatment outcomes.

Earlier studies have identified clinical manifestations during the progression of SFTS, such as bleeding manifestations, neurologic symptoms, persistent thrombocytopenia, and the occurrence of DIC and MOF, are the main risk factors



Expression profile of cytokines

Figure 9 Hierarchical cluster analysis of 67 cytokine related genes in healthy control, moderate, severe, and fatal groups. Each row represents mRNA and each column represents a sample. Red indicates higher expression and blue indicates low expression.

associated with mortality.³⁸ In addition to the immune landscape, the most distinctive feature observed in deceased patients in this study was coagulation dysfunction, specifically manifested as inhibition of platelet activation and coagulation, as well as a significant negative correlation between the gene modules that perform the relevant functions and mortality outcomes. Consistent with our results, previous studies have also found that the expression levels of platelet activation signaling pathway components were down-regulated during SFTSV infection with prolonged duration of infection, suggesting that SFTSV could induce thrombocytopenia by inhibiting platelet activation.³⁹ These research findings suggest that the inhibition of platelet activation function and the decrease in platelet count may synergistically induce coagulation dysfunction, increase the propensity for bleeding and DIC in SFTS patients, and lead to adverse outcomes. Therefore, in clinical treatment, especially for patients with bleeding manifestations, corresponding measures should be fully considered to improve the coagulation function of patients.

Conclusion

Taken together, our data revealed complex molecular response mechanisms underlying the clinical heterogeneity of SFTS patients and identified immunoregulatory phenomena and coagulation dysfunction features associated with lethal outcomes. These results extend our understanding of the pathogenesis of lethal SFTSV infection and contribute to the development of countermeasures against SFTSV. Nevertheless, the limitation of our study is that the sample size included in the study is small and limited to transcriptomic analysis, lacking further clinical and experimental data to validate potential mechanisms. In follow-up studies, larger clinical studies and more in-depth validation experiments need to be further carried out.

Data Sharing Statement

The transcriptomic data presented in the study was uploaded to the GEO database (<u>https://www.ncbi.nlm.nih.gov/geo/</u>) under accession number GSE264635.

Ethics Approval and Consent to Participate

All authors confirmed that this study complied with the Declaration of Helsinki. All participants signed written informed consent, and this study was approved by the Ethical Approval Committee of the Shandong Center for Disease Control and Prevention (Ethical approval number: SDJK-2021-25).

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Disclosure

The authors report no conflicts of interest in this work.

References

- 1. Yu XJ, Liang MF, Zhang SY, et al. Fever with thrombocytopenia associated with a novel bunyavirus in China. New Engl J Med. 2011;364 (16):1523-1532. doi:10.1056/NEJMoa1010095
- Li S, Li H, Zhang YL, et al. SFTSV infection induces BAK/BAX-dependent mitochondrial DNA release to trigger NLRP3 inflammasome activation. Cell Rep. 2020;30(13):4370–4385e4377. doi:10.1016/j.celrep.2020.02.105
- 3. Jiang XM, Xin QL, Liu K, et al. Regulation of the WNT-CTNNB1 signaling pathway by severe fever with thrombocytopenia syndrome virus in a cap-snatching manner. *mBio*. 2023;14(6):e0168823. doi:10.1128/mbio.01688-23
- 4. Hu YY, Zhuang L, Liu K, et al. Role of three tick species in the maintenance and transmission of severe fever with thrombocytopenia syndrome virus. *PLoS Negl Trop Dis*. 2020;14(6):e0008368. doi:10.1371/journal.pntd.0008368
- 5. Jiang XL, Zhang S, Jiang M, et al. A cluster of person-to-person transmission cases caused by SFTS virus in Penglai, China. *Clin Microbiol Infect.* 2015;21(3):274–279. doi:10.1016/j.cmi.2014.10.006
- 6. Jung IY, Choi W, Kim J, et al. Nosocomial person-to-person transmission of severe fever with thrombocytopenia syndrome. *Clin Microbiol Infect.* 2019;25(5):633e631–633e634. doi:10.1016/j.cmi.2019.01.006
- 7. Tran XC, Yun Y, Van An L, et al. Endemic severe fever with thrombocytopenia syndrome, Vietnam. *Emerg Infect Dis.* 2019;25(5):1029–1031. doi:10.3201/eid2505.181463
- 8. Win AM, Nguyen YTH, Kim Y, et al. Genotypic heterogeneity of orientia tsutsugamushi in scrub typhus patients and thrombocytopenia syndrome co-infection, Myanmar. *Emerg Infect Dis.* 2020;26(8):1878–1881. doi:10.3201/eid2608.200135
- Rattanakomol P, Khongwichit S, Linsuwanon P, Lee KH, Vongpunsawad S, Poovorawan Y. Severe Fever with Thrombocytopenia Syndrome Virus Infection, Thailand, 2019–2020. Emerg Infect Dis. 2022;28(12):2572–2574. doi:10.3201/eid2812.221183
- Wang M, Tan W, Li J, Fang L, Yue M. The endless wars: severe fever with thrombocytopenia syndrome virus, host immune and genetic factors. Front Cell Infect Microbiol. 2022;12:808098. doi:10.3389/fcimb.2022.808098
- 11. Peng C, Wang H, Zhang W, et al. Decreased monocyte subsets and TLR4-mediated functions in patients with acute severe fever with thrombocytopenia syndrome (SFTS). *Internat J Infect Dis.* 2016;43:37–42. doi:10.1016/j.ijid.2015.12.009
- 12. Sun Y, Jin C, Zhan F, et al. Host cytokine storm is associated with disease severity of severe fever with thrombocytopenia syndrome. *J Infect Dis.* 2012;206(7):1085–1094. doi:10.1093/infdis/jis452
- 13. Li MM, Zhang WJ, Liu J, et al. Dynamic changes in the immunological characteristics of T lymphocytes in surviving patients with severe fever with thrombocytopenia syndrome (SFTS). *Internat J Infect Dis.* 2018;70:72–80. doi:10.1016/j.ijid.2018.03.010
- 14. Wang G, Chang H, Jia B, et al. Nucleocapsid protein-specific IgM antibody responses in the disease progression of severe fever with thrombocytopenia syndrome. *Ticks Tick Borne Dis.* 2019;10(3):639–646. doi:10.1016/j.ttbdis.2019.02.003

- 15. Blanco-Melo D, Nilsson-Payant BE, Liu WC, et al. Imbalanced Host Response to SARS-CoV-2 Drives Development of COVID-19. *Cell*. 2020;181 (5):1036–1045e1039. doi:10.1016/j.cell.2020.04.026
- Lu S, Zhu N, Guo W, et al. RNA-Seq revealed a circular RNA-microRNA-mRNA regulatory network in hantaan virus infection. Front Cell Infect Microbiol. 2020;10:97. doi:10.3389/fcimb.2020.00097
- 17. Bermudez-Mendez E, Angelino P, van Keulen L, et al. Transcriptomic profiling reveals intense host-pathogen dispute compromising homeostasis during acute rift valley fever virus infection. J Virol. 2023;97(6):e0041523. doi:10.1128/jvi.00415-23
- Zhang L, Peng W, Wu J, et al. Pathogenicity and landscape of differential gene expression in mice orally infected with clinical coxsackievirus A6 (CA6). J Virol. 2024;98(2):e0135823. doi:10.1128/jvi.01358-23
- Zhang Y, Zhao L, Zhang J, et al. Antibody and transcription landscape in peripheral blood mononuclear cells of elderly adults over 70 years of age with third dose of COVID-19 BBIBP-CorV and ZF2001 booster vaccine. *Immu Age*. 2024;21(1):11. doi:10.1186/s12979-023-00408-x
- Robinson MD, McCarthy DJ, Smyth GK. edgeR: a Bioconductor package for differential expression analysis of digital gene expression data. *Bioinformatics*. 2010;26(1):139–140. doi:10.1093/bioinformatics/btp616
- 21. Ritchie ME, Phipson B, Wu D, et al. limma powers differential expression analyses for RNA-sequencing and microarray studies. *Nucleic Acids Res.* 2015;43(7):e47. doi:10.1093/nar/gkv007
- Szklarczyk D, Gable AL, Lyon D, et al. STRING v11: protein-protein association networks with increased coverage, supporting functional discovery in genome-wide experimental datasets. *Nucleic Acids Res.* 2019;47(D1):D607–D613. doi:10.1093/nar/gky1131
- Doncheva NT, Morris JH, Gorodkin J, Jensen LJ. Cytoscape StringApp: network analysis and visualization of proteomics data. J Prot Res. 2019;18 (2):623–632. doi:10.1021/acs.jproteome.8b00702
- 24. Wu T, Hu E, Xu S, et al. clusterProfiler 4.0: a universal enrichment tool for interpreting omics data. *Innovation*. 2021;2(3):100141. doi:10.1016/j. xinn.2021.100141
- 25. Langfelder P, Horvath S. WGCNA: an R package for weighted correlation network analysis. BMC Bioinf. 2008;9:559. doi:10.1186/1471-2105-9-559
- 26. Bindea G, Mlecnik B, Hackl H, et al. ClueGO: a Cytoscape plug-in to decipher functionally grouped gene ontology and pathway annotation networks. *Bioinformatics*. 2009;25(8):1091–1093. doi:10.1093/bioinformatics/btp101
- 27. Ding YP, Liang MF, Ye JB, et al. Prognostic value of clinical and immunological markers in acute phase of SFTS virus infection. *Clin Microbiol Infect*. 2014;20(11):0870–878. doi:10.1111/1469-0691.12636
- Kwon JS, Kim MC, Kim JY, et al. Kinetics of viral load and cytokines in severe fever with thrombocytopenia syndrome. J Clin Virol. 2018;101:57– 62. doi:10.1016/j.jcv.2018.01.017
- Casel MA, Park SJ, Choi YK. Severe fever with thrombocytopenia syndrome virus: emerging novel phlebovirus and their control strategy. *Exp Mol Med.* 2021;53(5):713–722. doi:10.1038/s12276-021-00610-1
- 30. Hu LF, Wu T, Wang B, et al. The regulation of seventeen inflammatory mediators are associated with patient outcomes in severe fever with thrombocytopenia syndrome. Sci Rep. 2018;8(1):159. doi:10.1038/s41598-017-18616-z
- 31. Kang SY, Yoo JR, Park Y, et al. Fatal outcome of severe fever with thrombocytopenia syndrome (SFTS) and severe and critical COVID-19 is associated with the hyperproduction of IL-10 and IL-6 and the low production of TGF-β. J Med Virol. 2023;95(7):e28894. doi:10.1002/jmv.28894
- 32. Yoo JR, Kim TJ, Heo ST, et al. IL-6 and IL-10 levels, rather than viral load and neutralizing antibody titers, determine the fate of patients with severe fever with thrombocytopenia syndrome virus infection in South Korea. *Front Immunol.* 2021;12:711847. doi:10.3389/fimmu.2021.711847
- 33. Liu J, Wang L, Feng Z, et al. Dynamic changes of laboratory parameters and peripheral blood lymphocyte subsets in severe fever with thrombocytopenia syndrome patients. *Internat J Infect Dis.* 2017;58:45–51. doi:10.1016/j.ijid.2017.02.017
- 34. Li MM, Zhang WJ, Weng XF, et al. CD4 T cell loss and Th2 and Th17 bias are associated with the severity of severe fever with thrombocytopenia syndrome (SFTS). *Clin Immunol.* 2018;195:8–17. doi:10.1016/j.clim.2018.07.009
- 35. Song P, Zheng N, Liu Y, et al. Deficient humoral responses and disrupted B-cell immunity are associated with fatal SFTSV infection. *Nat Commun.* 2018;9(1):3328. doi:10.1038/s41467-018-05746-9
- 36. Yamaoka S, Weisend C, Ebihara H. Identifying target cells for a tick-borne virus that causes fatal hemorrhagic fever. J Clin Invest. 2020;130 (2):598–600. doi:10.1172/JCI134512
- 37. Li H, Li X, Lv S, et al. Single-cell landscape of peripheral immune responses to fatal SFTS. Cell Rep. 2021;37(8):110039. doi:10.1016/j. celrep.2021.110039
- 38. Gai ZT, Zhang Y, Liang MF, et al. Clinical progress and risk factors for death in severe fever with thrombocytopenia syndrome patients. J Infect Dis. 2012;206(7):1095–1102. doi:10.1093/infdis/jis472
- 39. Fu H, Wang Y, Yuan C, et al. Screening core genes and signaling pathways after SFTSV infection by integrated transcriptome profiling analysis. *Virus Res.* 2023;332:199138. doi:10.1016/j.virusres.2023.199138

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