Severe Fever With Thrombocytopenia Syndrome in Southeastern China, 2011–2019

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Introduction: Severe fever with thrombocytopenia syndrome (SFTS) is an emerging infectious disease, and the number of cases has increased in recent years in Zhejiang Province, China. However, whether the seasonal distribution, geographic distribution, and demographic characteristics of SFTS have changed with the increase of incidence was unclear.

Materials and Methods: Data on SFTS cases in Zhejiang Province and tick density in Daishan County from 2011 to 2019 were collected. The changing epidemiological characteristics of SFTS including seasonal distribution, geographical distribution, and demographic features were analyzed using descriptive statistical methods, Global Moran’s I, local Getis-Ord Gi* statistic, and spatial scan statistic.

Results: A total of 463 SFTS cases including 53 (11.45%) deaths were reported from 2011 to 2019 in Zhejiang Province, and the annual number of cases showed increasing tendency. SFTS cases were reported in almost half of the counties (40/89) of Zhejiang Province. Elderly farmers accounted for most cases and the proportion of farmers has increased. Most cases (81.21%) occurred during April and August. The interval from illness onset to confirmation was significantly shortened (Z = 5.194, p < 0.001). The majority of cases were reported in Zhoushan City from 2011 to 2016, but most cases were reported in Taizhou City since 2017.

Discussion: We observed dynamic changes in the seasonal distribution, geographical distribution, and demographic features of SFTS, and comprehensive intervention measures, such as clearance of breeding sites, killing of tick adults, and health education should be strengthened in farmers of the key areas according to the changed epidemiological characteristics.

Keywords: severe fever with thrombocytopenia syndrome (SFTS), epidemiology, characteristic, spatiotemporal pattern analysis, tick

INTRODUCTION

Severe fever with thrombocytopenia syndrome (SFTS), caused by SFTS virus (SFTSV), was firstly confirmed in China in 2009 (1). SFTSV belongs to genus Phlebovirus in family Phenuiviridae, order Bunyavirales. In 2009, SFTS cases were firstly reported in Hubei Province and Henan Province, China. Subsequently, SFTS cases were also identified in South Korea, Japan, United Arab Emirates,
and Vietnam (2–5). In 2014, a new phlebovirus associated with severe febrile illness was identified in Missouri (6). Moreover, Zwiesel bat banyangvirus, a potentially zoonotic SFTSV-like banyangvirus was detectable in Northern bats from Germany (7). SFTSV has a wide range of vertebrate hosts, including domestic and wild animals, such as goats, sheep, cattle, dogs, pigs, and chickens (8, 9). Of note, direct infection of SFTSV from cats and dogs to humans has been reported and recognized as a social problem (10–13).

The main clinical manifestations of SFTS included high fever, gastrointestinal symptoms, and bleeding tendency. Most SFTS patients went through a self-limiting clinical course, but some cases developed serious symptoms (8, 14–17). Some of the patients (12–30%) may develop thrombocytopenia and leukopenia, and then die due to multiple organ failure (9, 14, 17–19). WHO had listed SFTS as one of the nine infectious diseases on its priority list, due to the increasing number of SFTS cases and its potential threat to human health.

Since 2010, more than 5,000 SFTS cases have been reported in more than 15 provinces of China (15). According to a survey of seroprevalence of SFTSV antibodies in Zhejiang Province, SFTS was endemic in rural areas of Zhejiang Province, and more than 400 SFTS cases have been reported since 2011 (20). With the development of detection technology, increasing awareness, and the change of environment factors, more and more SFTS cases were identified in Zhejiang Province. Here, we aim to summarize the changing epidemiological characteristics of SFTS in Zhejiang Province and analyze whether some characteristics have changed which would provide scientific information for its accurate control and prevention.

**MATERIALS AND METHODS**

**Case Definition**

A person with acute onset of fever (≥38.0°C) and other symptoms (e.g. gastrointestinal symptoms, bleeding), epidemiological risk factors (being a farmer or being exposed to ticks two weeks before illness onset) and laboratory data consisting of thrombocytopenia and leukocytopenia was defined as a suspected case. A suspected case with one or more of the following criteria: (1) detection of SFTSV RNA, (2) seroconversion or 4-fold increase in antibody titers between paired serum samples collected at least two weeks apart, and (3) isolation of SFTSV in cell culture was defined as a confirmed case.
FIGURE 2 | The number of severe fever with thrombocytopenia syndrome (SFTS) cases and case fatality rate in Zhejiang Province from 2011 to 2019.

FIGURE 3 | The median age and interval from illness onset to confirmation of SFTS cases in Zhejiang Province from 2011 to 2019. (A) is the median age and (B) is the interval from illness onset to confirmation.

Data Collection
Zhejiang Province, one of the southeastern coastal provinces of China, is located at 27°02′N to 31°11′N and 118°01′E to 123°10′E (Figure 1). The cities of Zhejiang Province and their subordinate counties are listed in Supplementary Table 1. The surveillance data of SFTS cases from January 2011 to December 2019 in Zhejiang Province were extracted from the communicable disease surveillance network system. We collected...
the following data: residential address, age, gender, occupation, date of illness onset, date of case confirmation, and the outcome (recovered or fatal). Furthermore, the demographic data of every county were collected from the statistical yearbook of Zhejiang Province.

**Ethics Statement**
This present research reported here has been approved by the Ethics Committee of Zhejiang Provincial CDC. Human research was carried out in accordance with the provisions of the Declaration of Helsinki.

**Data Analysis**
Means and $SD$ ($\bar{x} \pm s$) were applied if data conformed to normal distribution, otherwise median with interquartile range ($M \ [Q_1, Q_3]$) were used to describe the distribution of age, gender, occupation, duration onset to diagnosis annually. Enumeration data were described with frequencies and proportions annually. The Kruskal–Wallis test was used to compare the distribution...
of duration from illness onset to confirmation and the age distribution of SFTS cases. The chi-square test was used to compare SFTS's distribution of gender and occupation. The Jonckheere–Terpstra test, the Mann–Kendall trend test, and linear-by-linear association were used to conduct the trend analysis. 

ArcGIS 10.5 (ESRI, Redlands, CA, USA) was used to describe the geographical distribution of SFTS cases, and basic analysis was performed using R software version 4.0.0 (R Foundation for Statistical Computing, Vienna, Austria) and SPSS 18.0 (IBM Corp., Armonk, NY, USA).
Spatial Autocorrelation Analysis

Global Moran’s $I$ was used to explore the spatial pattern of the whole study area, the range of Global Moran’s $I$ is $[-1,1]$, and how closer its value is to 1. The closer the relationship between the spatial units is, the more similar their properties are, conversely, the closer it is to $-1$, the greater the difference between the spatial units (21). Z statistics and values were according to the results of Monte Carlo tests (999 times), and the main spatial weight matrix was set by the Queen method.

In addition, spatial heterogeneity is defined as the presence of significant variation in spatial autocorrelation within the study area. It is necessary to describe the variation of spatial autocorrelation in the study area. Therefore, we must rely on the local indexes that can detect local spatial autocorrelation to provide spatial autocorrelation values of regional units. Local indicators of spatial association were depicted by Local Getis-Ord $G_i^*$ statistic (22, 23). The expected value and the variance of the Local Getis-Ord $G_i^*$ statistic could be calculated by computer, so as to conduct the hypothesis test of whether the $G_i^*$ of the population is zero and deduce whether there are statistically significant high-value or low-value aggregation areas in the region and adjacent areas where the disease space $i$ and $j$ is distributed.

Space-Time Cluster Analysis

Spatial scan statistics was initially used to monitor the clusters of temporal aggregation. Additionally, the analysis was extended to three-dimensional space-time cluster by Kulldorff, which had been widely used in epidemiology and other fields. A cylindrical scanning window is established in the counties area of the research, and the bottom of the cylinder represents the size of the scanning area, and the height of the cylinder represents the length of the time range. The risk in each scanning window was examined, and the likelihood ratio (LLR) was also calculated. The value of LLR in different windows could be calculated and the value could be considered as statistics to compare and test by Monte-Carlo sampling. In our study, discrete Poisson distribution was selected to fit the probability model, high rates were set to scan for areas, the maximum spatial cluster size was reconfigured as 25% of the population at risk in the spatial window, the maximum temporal cluster size was given 10% of the study period (12 months), and the maximum number of

There are three categories on the $G_i^*$ Cluster Map: High-High (hot spots); Low-Low (cold spots); None (spatial outliers not significant). Spatial autocorrelation analysis was performed using GeoDa version 1.14.0.
replications was set as 999. The space-time cluster analysis was programmed in Software for the Spatial and Space-Time Scan Statistic (SaTScan version 9.3), and the result was performed using ArcGis 10.5 (ESRI, Redlands, CA, USA).

The Temporal and Spatial Distribution of Ticks in Daishan County
Daishan County is one of the most SFTS affected areas in Zhejiang Province, where ticks monitoring site was set up in 2013. There were three monitoring sites before 2017 and five monitoring sites after 2017 in Daishan County.

RESULTS
Epidemiological Characteristics
A total of 463 confirmed SFTS cases were reported from 2011 to 2019 in Zhejiang Province. The annual case was 9, 25, 31, 57, 72, 66, 71, 84, and 66 from 2011 to 2019 which showed an increasing tendency (Figure 2). The average case fatality rate (CFR) was 11.45% (53/463) and the difference of CFR in different years was not significant.

Of the 463 cases, 223 were men and 240 were women (the male-to-female ratio was about 0.929), and there was a significant difference ($\chi^2 = 15.192, p = 0.034$) across the different years. However, we found that the male-to-female ratio (ratio = 0.44) was particularly low in 2015, and the annual male-to-female ratio did not differ significantly ($\chi^2 = 4.601, p = 0.596$) across the different years (Supplementary Table 2).

The median age of all 463 cases was 66.00 (57.00, 74.00) and the annual median ages of cases from 2011 to 2019 were 70.00, 70.00, 63.00, 64.00, 65.00, 67.00, 68.00, 65.00, and 65.50, respectively (Figure 3A). The age distribution of different years was similar ($\chi^2 = 10.167, p = 0.254$). Of 463 SFTS cases, 321 cases (69.33%) were farmers accounting for the largest proportion and the proportion of farmers increased from 2011 to 2019 ($p < 0.001$).

From 2011 to 2019, SFTS cases were identified in 40 counties of Zhejiang Province. Of the 463 SFTS cases, 133 cases were reported in Daishan County, 71 cases in Linhui County, 51 cases in Tiantai County, 25 cases in Ninghai County, and 20 cases in Chun’an County. The numbers of SFTS affected counties from 2011 to 2019 were 4, 5, 7, 15, 13, 15, 20, 24, and 19, respectively (Figure 4).

The numbers of cases reported in Taizhou City, Zhoushan City, Ningbo City, Jinhua City, Hangzhou City, Shaoxing City, Huzhou City, Wenzhou City, and Lishui City were 158, 147, 54, 37, 27, 18, 15, 4, and 3, respectively. Notably, the numbers of SFTS cases were 6, 18, 21, 19, 31, 26, 14, 9, and 3 in Zhoushan City during 2011–2019, but the corresponding numbers were 1, 3, 0, 11, 32, 25, 33, 34, and 19 in Taizhou City. About 48.97% (95/194) of SFTS cases were reported in Zhoushan City from 2011 to 2015, but 41.26% (111/269) of the cases were reported in Taizhou City from 2016 to 2019 (Figure 5). The most seriously affected areas shifted from Zhoushan City to Taizhou City.

The heat map showed the monthly distribution of SFTS cases in Zhejiang Province from 2011 to 2019. A total of 376 cases (81.21%) occurred during April and August, and 426 cases (92.01%) occurred during April and October (Figure 6).

The median interval from illness onset to diagnosis of all cases was 6.00 (3.00, 9.00) days and the intervals from 2011 to 2019 were 12.00, 14.00, 7.00, 7.00, 5.00, 4.00, 5.00, 5.00, and 4.00 days, respectively. Of note, there was a significant difference (Kruska–Wallis Test $\chi^2 = 73.911, p < 0.001$) between the intervals in different years and a downtrend from 2011 to 2019 (Figure 3B).

Spatial Autocorrelation Analysis
The Moran’s I by global spatial autocorrelation analysis showed that positive global spatial correlations (Moran’s I = 0.491 in 2013, 0.431 in 2015, 0.401 in 2016, 0.513 in 2017, 0.187 in 2018, and 0.270 in 2019), which indicated that the positive correlation property of the regional distribution of SFTS cases (Table 1). The results of local spatial autocorrelation showed that the hot spot clusters (H–H, red color area) mainly located in some islands of southeastern coastal areas of Zhejiang Province, including Daishan, Shengsi, Dinghai, and Putuo, central-eastern areas of Zhejiang Province, including Tiantai, Ninghai, Linhai, Xiangshan, Sanmen, and Xianju from 2016, and western areas of Zhejiang Province, including Chun’an, Lin’an, and Anji from 2018 (Figure 7).

Space-Time Cluster Analysis
As summarized in Table 2, the most likely cluster mainly distributed in Zhoushan City (LLR = 29.228, $p = 1.200E-11 < 0.001$) including Daishan County, Shengsi County from 2015/4/27 to 2015/10/23. The secondary likely clusters are located in the east coast and northwest areas of Zhejiang Province. One cluster, including Xiangshan County, Ninghai County, Sanmen County, Tiantai County, Yinzhou County, Haishu County, Putuo County, Dinghai County (LLR = 8.037, $p = 0.021$) from 2017/4/16 to 2017/10/12, and the relative risk (RR) was 3.72. Additionally, another spatial-temporal cluster including Lin’an County, Anji County, Chun’an County, Zhiju County, Yuhang County, Jiande County, Pujiang County, Changxing County, Wuxing County, and Xiaoshan County from 2014/5/2 to 2014/10/28, and RR was 4.19. The other cluster included Lin’an County, Anji County, Chun’an County, Zhiju County, Yuhang County, Jiande County, Pujiang County, Changxing County,
The Temporal and Spatial Distribution of Ticks in Daishan County

Before 2017, the density of ticks ranged from 12.00 to 20.00. However, the density ranged from 9.20 to 12.00 in 2017, 2.20 to 4.13 in 2018, and 1.2 to 4.13 in 2019. The tick density was apparently on a downward trend.

DISCUSSION

In this study, we investigated the changing epidemiological characteristics of SFTS cases in Zhejiang Province from 2011 to 2019. Our study confirmed that there was an increasing tendency in the number of annual SFTS cases and affected counties. Several factors may be associated with the results. Since the identification of SFTS, more attention has been paid by health departments, and some measures have been conducted to improve the capacity of diagnosis in Zhejiang Province. For example, all municipal cities and counties where more than 5 cases had been reported established detection methods for SFTSV in Zhejiang Province. For example, all municipal cities and counties where more than 5 cases had been reported established detection methods for SFTSV in Zhejiang Province. The improvement of diagnosis and detection capacity may increase the number of reported SFTS cases. Similarly, the interval from illness onset to confirmation was significantly shortened from 2011 to 2019.

We also found that the medians of SFTS cases were all >60 years from 2011 to 2019, and the majority of the cases were farmers. The previous studies had reported that the elderly individuals might be more likely infected by SFTSV due to a decrease in immunity and underlying medical conditions (15, 16, 24). Another factor may also contribute to the result. In these affected counties, the majority of cases were elderly farmers living in hilly areas as young people went to cities to earn better money. Therefore, the elderly farmers had more chance of exposure to tick.

Similar to the results of other studies, most SFTS cases occurred from April to October. The seasonal distribution of SFTS cases might be related to tick dynamics. Ticks were believed as the carriers of SFTSV, and _Haemaphysalis longicornis_ was the predominant population carrying SFTSV (25, 26). The major habitats of _H. longicornis_ were the shrub, grass, and broad-leaved forest habitat. _H. longicornis_ were active in early March on vegetation, and there was an overlap among three stages with larvae active in late June. A number of _H. longicornis_ achieved peak in September, with nymphs active in early April and reaching peak in May and adults active from March to September and reaching peak in July (27–29). Moreover, the heat map of the monthly distribution of SFTS cases indicated that SFTS cases concentrated from April to October. The ecological characteristic of _H. longicornis_ was also in agreement with the corresponding results from the epidemiology of SFTS. Most ticks, especially those living in temperate and subtropical regions, had a distinct seasonal cycle in finding their hosts.

Of interest, most SFTS cases were reported in areas of Zhoushan City before 2016 and then the number of SFTS cases in Zhoushan City decreased sharply. The measures such as health education and risk communication, weed removal, livestock management, and tick control had been conducted since 2016 in Zhoushan City (24). Staff in Zhoushan City removed shrubs from residents’ surroundings, cleaned the lane around the house, and eradicated the weeds on both sides of the channel. Villagers cleared a width of at least 1 m above the no turf isolation belt and cleaned up the free ticks of the household environment and the ticks attached to raising animals. Health cleaning and insecticidal was implemented once a month for domestic animals and livestock (15, 24). As a result, the number of SFTS cases and tick density decreased in Zhoushan City. Our study also indicated that tick density decreased sharply in Daishan County. Other factors such as the SFTSV-carrying rate among ticks and molecular epidemiological characteristics of SFTSV may also contribute to SFTS incidence. To our disappointment, SFTSV was not detectable among ticks from Zhoushan City.
We used the global Moran’s I index to detect whether the positive spatial autocorrelation existed, and used Getis-Ord $G_i^*$ to explore the high-risk hotspots. We found hotspots and the gathering center was expanding from 1 center to 3 centers. Moreover, high-risk areas are mainly located at Daishan—Shengsi, Ninghai—Sanmen—Xiangshan—Linhai, and Anji—Lin’an—Chun’an. The heat map of the monthly distribution of SFTS cases indicated that SFTS cases concentrated from April to October, which was accordant with the gathering time of three spatial-temporal clusters. Of note, the time frame of clusters indicated that the center of a cluster of SFTS was not only transferring but also expanding.

There are several limitations to the study. First, data of SFTS cases were collected from a passive surveillance system. Some factors such as detection capability and availability of health facilities may influence the data quality. Second, other factors may also be associated with SFTS incidence. For example, the SFTSV-carrying rate of ticks, molecular epidemiological characteristics of SFTSV, and the meteorological factor may be associated with SFTS occurrence. But we could not get data on these factors in the study, and additional research is warranted to elucidate the factors associated with SFTS incidence.

In summary, our study analyzed the changing epidemiological characteristics of SFTS in Zhejiang Province. We found that the annual number of SFTS cases and affected counties showed an increasing tendency in Zhejiang Province. Some epidemiological characteristics have changed. For example, high incidence areas transferred from Zhoushan City to Taizhou City, the proportion of farmers increased year by year, and the interval from illness onset to confirmation decreased year by year. Comprehensive intervention measures such as clearance of breeding sites, kill of tick adults, and...
health education should be adjusted according to the changed epidemiological characteristics.

DATA AVAILABILITY STATEMENT
The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

AUTHOR CONTRIBUTIONS
MT is responsible for conceptualization, methodology, software, and data curation. YC is responsible for software and visualization. RZ, JR, XS, and SG are responsible for data curation and supervision. YL, JS, and JJ are responsible for conceptualization, methodology, software, data curation, writing, reviewing and editing. All authors contributed to the article and approved the submitted version.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fpubh.2021.803660/full#supplementary-material

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