- 1 Article
- 2 Using Multiple Index Comprehensive Method to
- 3 Assess Urban Rainstorm Disasters Risk in Jiangsu
- 4 Province, China
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- Abstract: A new model for risk assessment of urban rainstorm disasters, based on information diffusion method and variable fuzzy sets (IDM-VFS) was proposed. In addition, an integrated index system of urban rainstorm risk was established. In the proposed model, IDM was employed to calculate the classification standards of urban rainstorm risk levels, then the VFS was adopted to assess the dangerousness, sensitivity, vulnerability and comprehensive risk of urban rainstorm disasters. In the present study, the urban rainstorm risk of Jiangsu province was evaluated with the proposed model. The results show that Wuxi, Changzhou, Nanjing and Suzhou have higher dangerousness, due to sustained rainfall and strong rainfall intensity in short duration; Wuxi, Changzhou and Nanjing have higher sensitivity because of lower disaster resistance ability; and Wuxi and Suzhou have higher vulnerability because these cities have higher potential losses in face of urban rainstorm disasters. The comprehensive risk zoning map of urban rainstorm shows apparent regional characteristics: the northwestern cities have lower risk than the southern cities. Moreover, most cities of the Jiangsu province are of the moderate urban rainstorm risk level. The results are consistent with the actual situation of Jiangsu province, and the study can provide some decision-making references for the urban rainstorm management.
- **Keywords:** urban rainstorm disasters; risk assessment; information diffusion method (IDM); variable fuzzy sets (VFS); Jiangsu province

### 1. Introduction

With the global climate change and the rapid developments of the urbanization, many cities are suffered extreme rainstorm events frequently [1]. About 15% of the global population is under the threat of the rainstorm disasters as a consequence of combined impacts of climate change phenomenon and rapid population growth, also the rainstorm events tend to increase both in frequency and magnitude [2-4]. According to EM-DAT and Munich RE disaster databases, flood losses account for 28% of the total global disaster losses. Now the urban rainstorm disasters occur more frequently in big cities in the world, which causes great impacts on urban traffic safety and brings about severe economic losses [5,6]. In China, more than 600 cities are exposed to frequent flooding, which has a huge social, economic and environmental impact [7]. Frequent urban rainstorms have severely restricted the stability of China's urban developments [8]. Therefore, it is important to assess the risk of urban rainstorm disasters, which can help to improve the prediction ability and to reduce the losses caused by urban rainstorm disasters.

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2 of 14

In order to assess the risk of urban rainstorm disasters, it need build a reasonable and effective assessment system. Over the years, several disaster risk assessment systems of urban rainstorm have been developed. Urban rainstorm is a multivariate phenomenon which is always assessed by many integrated indices, including economic loss data, social impact and emergency performance, etc [9-11]. Lyu et al. [12] selected rainy season, average rainfall, average rainy days, elevation and slope, river proximity, river density, land use, metro line proximity, metro line density and road network proximity and density to construct a risk assessment system of metro systems, and took Guangzhou metro system as an example to conduct an empirical study. Alfa [13] proposed a flood risk assessment of Ofu river catchment in Nigeria, including elevation, slope, proximity and soil type. Weerasinghe et al [14] put forward the risk assessment system of rainstorm for the Western Province of Sri Lanka, and use a statistical expression of hazard, exposure and vulnerability to assess the combined flood risk levels. These researches laid a good foundation for the risk assessment index system of urban rainstorm disasters. Therefore, in order to assess the risk of urban rainstorm disasters effectively, this paper establishes the urban rainstorm assessment system with dangerousness indices of hazardformative factors, sensitivity indices of hazard-inducing environments, and vulnerability indices of hazard-affected bodies.

Then in order to assess the risk of urban rainstorm disasters properly, it is necessary to adopt scientific and effective methods. Many studies have focused on hydrological and meteorological models or traditional assessment methods to assess the risk of urban rainstorm disasters. For example, Jonkman et al. [15] proposed a model for the estimation of damage caused by rainstorm in the Netherlands. This integrative character of the presented model is featured by the combination of information on land use and economic data, and data on flood characteristics and stage-damage functions. Wang et al. [16] developed a generalized risk assessment model of pipeline network, in combination with critical data of urban plan, as well as hydrologic data, and used Storm Water Management Model (SWMM) to optimally analyze on the rate of impervious area of upper catchments areas as well as the design standard of lower drainage pipeline. Shao et al. [17] used the methodology of gray fixed weight cluster analysis to assess the disaster losses from 2004 to 2009 based on the 1302 historical sample data. Some advanced technologies such as GIS (Geographic Information System) and remote-sensing imagery [18-21] are used for risk assessment of urban rainstorm disasters. While urban rainstorm disasters are difficult to quantify accurately mainly because of the lack of sufficient data<sup>[22]</sup>. So fuzzy and uncertainty theories can be introduced into the risk assessment of urban rainstorm disasters, such as fuzzy comprehensive evaluation [23], information diffusion [24], variable fuzzy sets (VFS) [25]. However, there are some drawbacks in these methods, for example, the fuzzy comprehensive evaluation cannot well resolve the duplication information caused by related evaluation indices; the establishment of the relative membership function of the variable fuzzy sets depends on physical analysis and experts' judgments [26]. Hence, taking these factors into consideration, it can combine the advantages of different methods to construct a new approach.

Urban rainstorm disasters lack enough data to assess, while the information diffusion method (IDM) can compensate for the incomplete information of small samples. And the risk of urban rainstorm disasters is ambiguous and uncertain, it have many different indices to calculate. So it can use the variable fuzzy sets (VFS) to eliminate the border effect on assessment results effectively and make full use of various indices data to obtain comprehensive risk evaluation results. In this paper, IDM was combined with VFS as an integrated model to assess the risk of urban rainstorm disasters. The IDM is employed to calculate the level classification standards of assessment indices, and then the VFS is adopted to assess the comprehensive risk of urban rainstorm. Through integrated assessment system and IDM-VFS model, Jiangsu province in China can be taken as a case study area. The paper is organized as follows. Section 2 describes the overview of study area, then the data and methods are followed in Section 3. Section 4 depicts the main results and analysis. The discussion and conclusions of the study are presented in Section 5.

## 2. Overview of Study Area

This paper takes Jiangsu province as the study area. Jiangsu province (Figure 1), which is located between 116°18′E-121°57E and 30°45′N-35°20′N, the total land area is 1.072×10<sup>5</sup> km², accounting for 1.1% of the total area of China. The geomorphology of Jiangsu is mainly plains, covering 68% of the total area. The elevation of most area is lower than 50m [27]. There are 13 prefecture-level cities in Jiangsu province, including Nanjing (provincial capital city), Wuxi, Xuzhou, Changzhou, Suzhou, Nantong, Lianyungang, Huai'an, Yancheng, Yangzhou, Zhenjiang, Taizhou and Suqian.

Jiangsu province has become one of the most urbanized regions, all the 13 cities in Jiangsu province are prosperous cities in China. In recent years, urban rainstorm disasters have occurred frequently and caused huge losses in each city of Jiangsu province. Thus, urban rainstorm has become one of the most important factors that restrict the development of Jiangsu province.

Jiangsu province belongs to the subtropical monsoon climate, it is usually hot and rainy in summer, and most of the large-scale precipitation in Jiangsu province is in summer. Thus, in order to better show the actual situations of the urban rainstorm disasters of Jiangsu province, this paper emphasizes on the period which is from June to August.

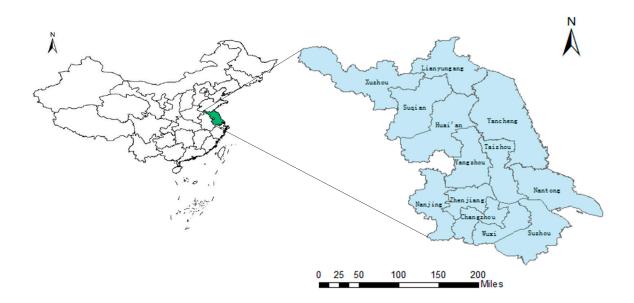


Figure 1. Geographic location of Jiangsu Province, China.

### 3. Data and Methods

# 19 3.1. Data Sources

This paper uses Jiangsu Statistical Yearbook data and China City Statistical Yearbook data from the National Bureau of Statistics of China to collect the statistical information of all cities in Jiangsu province including economic, social, demographic,, urban construction, environmental and other related city statistics. The meteorological data and the rainfall statistics are collected from the meteorological stations of cities in Jiangsu province. Some other historical precipitation statistics are provided from the Jiangsu climate center.

### 3.2. Weight of Index System

The risk assessment index system of urban rainstorm disaster includes a lot of indices. The index weight reflects the relative importance of each index in the risk assessment index system. In this paper, the AHP (Analytic Hierarchy Process) and entropy weight method are combined to calculate

the weights of indices. The AHP is a subjective method to determine the weight of indices based on experts' experience [28]. The entropy weight method is an objective method that determines the importance of index based on the data information [29]. This combined method can reduce the interference of the individual's subjective judgments on the weights of indices. Therefore, it can make the weights more realistic and reliable relatively [30]. The idea of calculating weights is to calculate the subjective weights of indices by AHP, and then to establish the entropy model of expert's own weights by using the subjective weights of indices as attribute matrices through the idea of information entropy. The weights of the indices are corrected by the weights of the experts, and the final combination weights are obtained.

Suppose the weights calculated by AHP are the subjective weights of indices. The subjective weights of indices can be represented as follows:

$$W_z = (w_{z1}, w_{z2}, \dots, w_{zr})^T$$
 (1)

- 12 where the  $W_z$  is the subjective weight vector, then the  $0 < w_{zs} < 1$  and  $\sum_{s=1}^{r} w_{zs} = 1$  (
- 13  $s = 1, 2, \dots, r; z = 1, 2, \dots, v$ ).

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Then the expert's own weights are calculated by entropy weight method, it can be shown as follows:

$$S = (S_1, S_2, \dots, S_v)^T$$
 (2)

- where the S is expert's own weight vector and, and  $0 < S_z < 1$  ( $z = 1, 2, \dots, v$ ) and  $\sum_{z=1}^{v} S_z = 1$ .
- Finally, the weight fusion vector of subjective weights and expert's own weights can be expressed as  $W = (w_1, w_2, \dots, w_r)^T$ , and it can be calculated as follows:

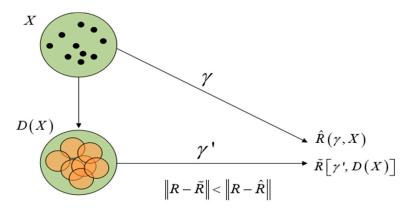
$$w_s = \sum_{z=1}^{\nu} \left( w_{zs} \times S_z \right) \tag{3}$$

- 19 where  $0 < w_s < 1$ ,  $\sum_{s=1}^{r} w_s = 1$  and  $s = 1, 2, \dots, r; z = 1, 2, \dots, v$ .
- 20 3.3. Information Diffusion Method (IDM)

This paper uses normal information diffusion method, IDM is a kind of fuzzy mathematical processing method which can be used to optimize the fuzzy information of samples by means of appropriate diffusion model [31]. When the sample is incomplete, the method can make the diffusion estimation closer to the real relationship than the non-diffusion estimation. This method can establish the level classification standards of the urban rainstorm disasters assessment indices to improve the evaluation accuracy [32]. Based on the historical data of Jiangsu province, the different risk levels of indices were determined according to classification standards.

The principle of information diffusion method can be expressed as follows: suppose X (  $X = \{x_1, x_2, \dots, x_n\}$ ) is a given sample, it can be used to estimate a relationship on a domain U.

- $A = \{x_1, x_2, \dots, x_n\}$  ) is a given sample, it can be used to estimate a relationship on a domain C.
- 30 When the X is incomplete, there exists an appropriate diffusion function  $f(x_i,u)$  and the
- 31 corresponding operator  $\gamma'$  , which can transform X into a fuzzy sample D(X) , so that the
- 32 information with a value of 1 from the sample  $\, X \,$  can be diffused around the sample following the
- 33 function  $f(x_i,u)$ , and the diffusion estimate  $\tilde{R}$  is closer to the real relationship than the non-
- 34 diffusion estimate  $\hat{R}$  [33-35]. It can be shown in the Figure 2.



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**Figure 2.** The principle of information diffusion method.

3 The calculation steps of the information diffusion are shown as follows:

Firstly, the risk levels of urban rainstorm disasters are divided into five levels, that is lowest risk, lower risk, moderate risk, higher risk and highest risk. The information carried by  $x_i$  ( $i=1,2,\cdots,n$ ) which can be chosen from the sample X can be diffused into  $u_j$  ( $j=1,2,\cdots,m$ ) from the domain U according to Equation (4).

$$f_i(u_j) = \frac{1}{h\sqrt{2\pi}} \exp\left[-\frac{(x_i - u_j)}{2h^2}\right]$$
 (4)

8 where h is the diffusion coefficient which can be calculated by different sizes and values of the samples.

Secondly, in order to make each set of sample values identical, the diffusion function  $f_i(u_j)$  is normalized to be the diffusion function  $g_i(u_j)$ :

$$g_i(u_j) = \frac{f_i(u_j)}{\sum_{i=1}^m f_i(u_j)}$$
(5)

So the probability of samples located in  $u_i$  can be represented as follows:

$$F(u_{j}) = \frac{\sum_{i=1}^{n} g_{i}(u_{j})}{\sum_{j=1}^{m} \sum_{i=1}^{n} g_{i}(u_{j})}$$
(6)

Finally, the exceeding probability  $F^*(u_j)$  can be obtained through Equation (7):

$$F^*(u_j) = \sum_{k=j}^m F(u_k) \tag{7}$$

According to the classification standards of exceeding probability, the critical value of each evaluation index corresponding to the risk level of urban rainstorm is obtained, so the level classification standards of urban rainstorm risk assessment indices can be obtained.

### 3.4. Variable Fuzzy Set (VFS)

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Variable fuzzy set theory is mainly used in the dynamic analysis of fuzzy phenomena in the system [36]. As its core, relative membership function, relative difference function and variable fuzzy set quantify the process of changing things from quantity to quality and describe it with mathematical

languages. By changing model and model parameters, the credibility and reliability of evaluation, identification and decision-making can be increased, which provides new ideas for risk assessments in many fields [37, 38].

Fuzzy variable evaluation method calculates the evaluation level of urban rainstorm disasters scientifically by changing the model and its parameter combination, and it can improve the reliability of risk assessment results. The fuzzy variable evaluation method mainly includes the following steps:

- (1) Generating index eigenvalue matrix
- Suppose there is a sample set  $Y = \{y_1, y_2, \dots, y_n\}$  consisting of n samples of natural disasters.
- The index eigenvalue of the sample i can be expressed as  $y_i = (y_{1i}, y_{2i}, \dots, y_{ri})^T$  where r is the number of sample indices. Then the sample set can be express as  $Y = (y_{si})_{r \times n}$ , where
- 11  $s = 1, 2, \dots, r; i = 1, 2, \dots, n$ .

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(2) Establishing index standard eigenvalue matrix

Suppose there are *m* levels of assessment classification standards. And sample set is identified according to different standards of eigenvalues of *r* sample indices, and then the standard eigenvalue matrix of the first order index is obtained.

- (3) Calculating the relative membership matrix of index level
- The interval matrix and the bound matrix of variable set of indices can be determined by referring to the standard value matrix of indices and the actual situation of the area. Then according to the different eigenvalues of samples  $y_{si}$ , the different degree  $L_A(y_{si})$ , so relative membership degree matrix can be calculated as follows:

$$\mu_{A}(y_{si}) = \frac{1 + L_{A}(y_{si})}{2} \tag{8}$$

- (4) Determining the weight of each index and the relative membership degree
- According to the Equation (8), the non-normalized relative membership degree can be calculated as follows:

$$T_{hi}^* = \left\{ 1 + \left[ \frac{\sum_{s=1}^r \left[ w_s (1 - \mu_A(y_{si})) \right]}{\sum_{s=1}^r \left[ w_s \mu_A(y_{si}) \right]^p} \right]^{\frac{\alpha}{p}} \right\}^{-1}$$
(9)

- 24 where  $w_s(s=1,2,\cdots,r)$  is the index weight that can be calculated by Equation (3). And r is the
- identify number of indices, h is the risk level number where  $h = 1, 2, \dots, m$ ,  $\alpha$  is the optimal rule
- parameter  $(\alpha = 1, 2)$ , P is the distance parameter, P = 1 is Hamming distance and P = 2 is
- 27 Euclidean distance. Then the normalized relative membership degree can be calculated as follows:

$$T_{h_i} = \frac{T_{h_i}^*}{\sum_{h=1}^m T_{h_i}^*} \tag{10}$$

- Finally, according to the principle of the largest degree of membership, we can obtain the risk levels of urban rainstorm disasters.
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## 4. Results and Analysis

### 4.1. Risk Index System of Urban Rainstorm

The reasonable assessment indices of urban rainstorm disasters should be able to accurately describe the risk of urban rainstorm disasters. Therefore, In this paper, an integrated risk assessment index system of urban rainstorm disasters was established (see Table 1). The index system is divided into three subsystems including dangerousness, sensitivity and vulnerability.

The dangerousness indices reflect the abnormal conditions and factors of external natural environment. The risk of urban rainstorm disasters can be attributed to short-term rainfall far exceeding normal situations or long-term rainfall in cities, which will lead to the arranged discharge of rainwater beyond the capacity of urban drainage network. Generally, the larger the dangerousness is, the higher the risk of urban rainstorm disasters is. This paper chooses continuous rainfall days ( $I_{11}$ : days), heavy rain days ( $I_{12}$ : days), maximum rainfall in 24h ( $I_{13}$ : mm), monthly total rainfall ( $I_{14}$ : mm) and precipitation anomaly percentage ( $I_{15}$ : %) as the evaluation indices of dangerousness.

The sensitivity indices represent that a particular region is potential to the destruction and influence of disasters due to various natural and social factors [39]. Jiangsu province is located in the plain area with low altitude, and it is easy to cause floods once it encounters rainstorms. And rapid urbanization leads to the change of urban surface attributes. The urban surface is mostly impervious and hardened surface, which leads to the rapid convergence of surface rainwater under extreme rainstorms. The construction of drainage pipeline network in cities also has not kept pace with the development of the city. So the urban average elevation ( $I_{21}$ : m), urban green coverage rate ( $I_{22}$ : %), urban drainage network density ( $I_{23}$ : km/km²), urban water area percentage ( $I_{24}$ : %) and impermeable construction land ( $I_{25}$ : km²) were selected as the sensitivity indices.

The vulnerability indices describe the potential losses of the area exposed to the risk<sup>[40]</sup>. It refers to the possible impact of urban rainstorm disasters in the urban, that is, the level of loss caused by urban rainstorm. It is generally believed that densely populated, industrially developed cities suffer greater risks and losses in face of the urban rainstorm disasters. The vulnerability indices include the density of affected population ( $I_{31}$ : people/km<sup>2</sup>), GDP of unit area ( $I_{32}$ :100 million yuan/km<sup>2</sup>), disaster relief investment level ( $I_{33}$ :%) and public emergency response capability ( $I_{34}$ ).

Table 1. Risk assessment index system of urban rainstorm disaster.

Target layer	Primary indices	Secondary indices	
Urban Rainstorm Disaster Risk		Continuous Rainfall Days(I11:days)	
	Dangerousness	Heavy Rain Days (I12:days)	
		Maximum Rainfall in 24h (I13:mm)	
		Monthly Total Rainfall (I14:mm)	
		Precipitation Anomaly Percentage (I15:%)	
	Sensitivity	Urban Average Elevation (I21:m)	
		Urban Green Coverage Rate (I22:%)	
		Urban Drainage Network Density (I23:km/km²)	
		Urban Water Area Percentage (I24:%)	
		Impermeable Construction Land (I25:km²)	
	Vulnerability	Density of Affected Population (I31:people/km²)	
		GDP of Unit Area (I32:100 million yuan/km²)	
		Disaster Relief Investment Level (I33:%)	
		Public Emergency Response Capability (I34)	

#### 4.2. Risk Evaluation Based on IDM-VFS Model

Based on the risk assessment index system and IDM-VFS model, firstly the AHP was combined with entropy method to determine the weights of the risk indices of urban rainstorm disasters; then the IDM was adopted to determine the classification standards of the risk indices; and the disaster risk values in dangerousness, sensitivity and vulnerability can be calculated by the VFS model respectively. Finally, the comprehensive disaster risk levels were obtained and the risk zoning map was drawn.

# 4.2.1. Determine the Weights of Risk Indices

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Indices weights are determined by combined AHP and entropy weight method. The weights of risk assessment indices are shown in Table 2.

**Table 2.** The index weights of urban rainstorm disasters.

Primary indices	Secondary indices	Weight $\mathcal{O}_i$
	Continuous Rainfall Days(days)	0.0630
Dangerousness	Heavy Rain Days (days)	0.0735
	Maximum Rainfall in 24h (mm)	0.0945
	Monthly Total Rainfall (mm)	0.0665
	Precipitation Anomaly Percentage (%)	0.0425
Sensitivity	Urban Average Elevation (m)	0.0772
	Urban Green Coverage Rate (%)	0.0577
	Urban Drainage Network Density (km/km²)	0.0927
	Urban Water Area Percentage (%)	0.0735
	Impermeable Construction Land (km²)	0.1279
Vulnerability	Density of Affected Population (people/km²)	0.0424
	GDP of Unit Area (100 million yuan/km²)	0.0916
	Disaster Relief Investment Level (%)	0.0452
	Public Emergency Response Capability	0.0408

#### 4.2.2. Calculate the Level Classification Standards of Indices

The level classification standards of each risk assessment index of urban rainstorm disasters are determined by IDM. Firstly, the index values can be taken as samples of information diffusion, then the exceeding probability of each index also can be calculated. Finally, the level classification standards of each risk assessment secondary indices are obtained (see Table 3).

Table 3. Level classification standards of risk assessments secondary indices.

	First	Second	Third	Forth	Fifth
Secondary Indices	Level	Level	Level	Level	Level
	Lowest	Lower	Moderate	Higher	Highest
Continuous Rainfall Days(days)	<1	1~2	2~4	4~6	>6
Heavy Rain Days (days)	<1	1~3	3~5	5~7	>7
Maximum Rainfall in 24h (mm)	<25	25~50	50~100	100~200	>200
Monthly Total Rainfall (mm)	< 50	50~124	124~236	236~378	>378
Precipitation Anomaly Percentage(%)	<4	4~15	15~40	40~100	>100
Urban Average Elevation (m)	>35	35~20	20~10	10~5	<5
Urban Green Coverage Rate (%)	>50	50~40	40~30	30~20	<20
Urban Drainage Network Density (km/km²)	>32	32~24	24~16	16~10	<10
Urban Water Area Percentage (%)	>30	30~20	20~15	15~10	<10
Impermeable Construction Land (km²)	<90	90~148	148~245	245~440	>440
Density of Affected Population (people/km²)	<1265	1265~2355	2355~3375	3375~4430	>4430
GDP of Unit Area(100 million yuan/km²)	< 0.8	0.8~1.2	1.2~3	3~5	>5
Disaster Relief Investment Level (%)	>17	15~17	13~15	9~13	<9
Public Emergency Response Capability	>90	90~80	80~70	70~60	<60

### 4.2.3. Calculate the Disaster Risks of Three Subsystems

According to Table 3, the interval matrix, bound matrix are established, and then, based on Equations (8)-(10), the relative membership degree matrix and integrated membership degree are obtained, and finally the risk values can be calculated in terms of dangerousness, sensitivity and vulnerability respectively. For demonstration purposes, Nanjing has been chosen as an example to discuss the risk assessment of urban rainstorm disasters in detail. The risk values of urban rainstorm disasters in Nanjing in 2016 are shown in Table 4.

Table 4. Risk values from June to August in 2016 in Nanjing.

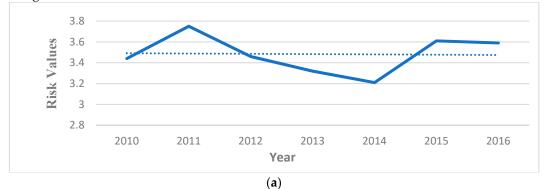
Month	Dangerousness	Sensitivity	Vulnerability
June	4.11	3.82	2.87
July	4.21	3.82	2.85
August	2.46	3.82	2.88
Average	3.59	3.82	2.87

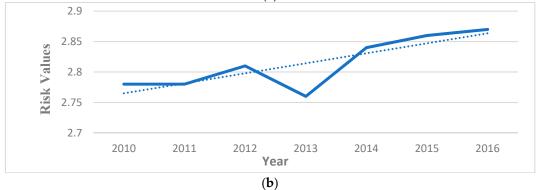
As we can see from the Table 4, dangerousness indices have different risk values due to different monthly rainfall and rainfall days. While sensitivity and vulnerability indices remain basically unchanged in one year. Adopting the same methods and procedures, we can obtain risk values of the urban rainstorm in Nanjing from 2010 to 2016 in terms of dangerousness, sensitivity and vulnerability respectively (see Table 5).

Table 5. Risk values of urban rainstorm from 2010 to 2016 in Nanjing.

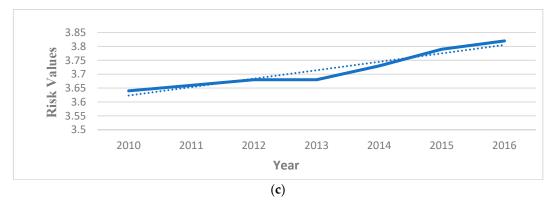
Year	Dangerousness	Sensitivity	Vulnerability
2010	3.44	3.64	2.78
2011	3.75	3.66	2.78
2012	3.46	3.68	2.81
2013	3.32	3.68	2.76
2014	3.21	3.73	2.84
2015	3.61	3.79	2.86
2016	3.59	3.82	2.87
Average	3.48	3.72	2.82

From the Table 5, the average risk value of dangerousness is 3.48, the average risk value of sensitivity is 3.72, and the average risk value of vulnerability is 2.82. The dangerousness of 2011 is higher than other years because the precipitation was higher than the normal level (Figure 3(a)), it had sustained rainfall and strong rainfall intensity in short duration. While the sensitivity and vulnerability had an upward tendency because of the acceleration of urbanization (Figure 3(b),(c)). The impervious construction area of the cities is gradually increasing, but the drainage facilities and greening constructions are not matched, and the density of affected population and GDP were also increasing from 2010 to 2016.





10 of 14



**Figure 3.** Variation tendency of risk values in terms of three subsystems from 2010 to 2016. **(a)** dangerousness; **(b)** sensitivity; **(c)** vulnerability.

By calculating the average risk values in different cities in terms of dangerousness, sensitivity and vulnerability, and weighting the three subsystems with weights, the risk level of urban rainstorm disasters can be shown in Table 6.

Table 6. Comprehensive risk level in Jiangsu province.

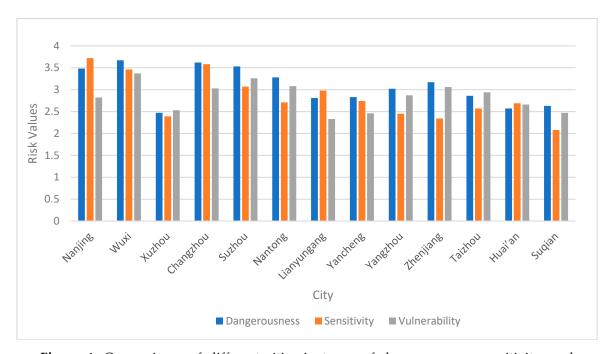
City	Dangerousness	Sensitivity	Vulnerability	Risk level
Nanjing	3.48	3.72	2.82	4
Wuxi	3.67	3.46	3.37	4
Xuzhou	2.47	2.39	2.53	2
Changzhou	3.62	3.58	3.03	4
Suzhou	3.53	3.07	3.26	3
Nantong	3.28	2.71	3.08	3
Lianyungang	2.81	2.98	2.33	3
Yancheng	2.83	2.74	2.46	3
Yangzhou	3.02	2.45	2.87	3
Zhenjiang	3.17	2.34	3.06	3
Taizhou	2.86	2.57	2.94	3
Huai'an	2.57	2.69	2.66	3
Suqian	2.63	2.08	2.47	2

The dangerousness of Wuxi, Changzhou, Nanjing and Suzhou are higher, while Xuzhou, Huai'an and Suqian are lower from 2010 to 2016. The major influence factors of dangerousness are sustained rainfall and strong rainfall intensity in short duration. And the precipitation decreased from south to north gradually.

The sensitivity of Wuxi, Changzhou and Nanjing are higher, while Xuzhou and Suqian are lower. The sensitivity of urban rainstorm disasters mainly depends on the natural and social environment of the cities and the disaster resistance level. The conditions of the different cities in Jiangsu province are uneven. Because different cities have different natural and social environments, Zhenjiang, Xuzhou and Suqian have higher altitudes and less impervious construction area which makes them have lower sensitivity. Wuxi, Changzhou and Nanjing are more advanced so they have more impervious construction area which decreased the ability of disaster resistance, resulting in higher sensitivity.

Lianyungang, Yancheng and Suqian are located in the lowest vulnerability area, while Wuxi and Suzhou are the highest cities. The vulnerability of urban rainstorm is the reflection of the vulnerable degree of social economy and human society capability to disasters. Lianyungang, Yancheng and Suqian have lower GDP of unit area and affected population density so they belong to the lower disaster vulnerability cities. The GDP of unit area in Wuxi and Suzhou is more than 500 million yuan/ $km^2$  and the population density is higher, leading to the highest urban rainstorm vulnerability.

The comparisons of different cities in terms of dangerousness, sensitivity, and vulnerability respectively in Jiangsu province were shown in Figure 4.



**Figure 4.** Comparisons of different cities in terms of dangerousness, sensitivity, and vulnerability respectively in Jiangsu province.

Based on the assessment results, the comprehensive risk zoning map in Jiangsu province can be drawn (Figure 5). From the Figure 5, it can be seen that the comprehensive risks of urban rainstorm in Jiangsu province have apparent regional characteristics. The comprehensive risk level of urban rainstorm in Wuxi, Changzhou and Nanjing is higher, while Xuzhou and Suqian is lower. The risk levels in the northwestern cities are lower than the southern cities in the Jiangsu province. Moreover, most cities of Jiangsu province experience moderate risk level.

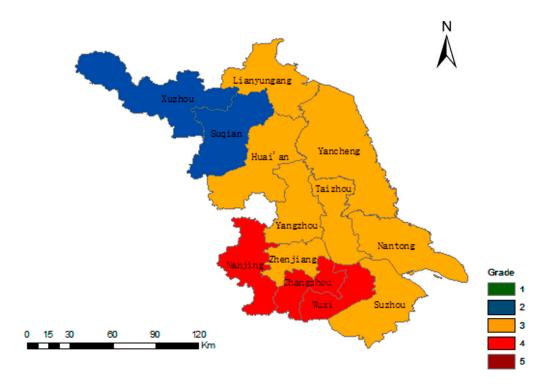


Figure 5. Distribution of urban rainstorm risk in Jiangsu province.

Because the urban rainstorm disasters are occurring more frequently in China, so the risk assessment of urban rainstorm disasters is very important for urban rainstorm management. However, urban rainstorm is a fuzzy phenomenon and there is a lack of sufficient data. It is difficult to depict it accurately through traditional statistics, especially in small sample problems. Therefore, the method built in this paper can be developed to assess the risk of urban rainstorm disasters. The VFS can make full use of various index data to obtain comprehensive risk evaluation results. The IDM can transform a sample observed value into a fuzzy set and is capable of dealing with small sample problems. Meanwhile, IDM can extract useful information and establish the level classification standards of the risk assessment indices, which is helpful to determine the relative membership function of VFS.

According to the natural disaster system theory and actual situation of Jiangsu province, an integrated risk assessment system of urban rainstorm disasters with dangerousness indices, the sensitivity indices and the vulnerability indices are established. Considering the integrated impact of dangerousness, sensitivity and vulnerability indices, to establish an integrated risk index system of urban rainstorm disasters in Jiangsu province is helpful to support risk management and develop effective mitigation strategies in this region. In addition, most cities of Jiangsu province are located in a moderate risk region. The results are consistent with the actual situation of the Jiangsu province and can provide reference and decision-making for management departments of rainstorm disasters. However, the assessment of urban rainstorm disasters is quietly complex. It needs to be studied further in formation mechanism, risk factors and losses of disasters.

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- **Author Contributions:** J.C. had the original idea for the study; M.C. and P.Z. were responsible for data collection, M.C. and J.C. analyzed the data and wrote the paper. P.Z. provided guidance, comments and key suggestions;
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