

# Risk Zoning of Typhoon Disasters in Zhejiang Province, China

Yi Lu<sup>1</sup> (陆逸), Fumin Ren<sup>2</sup> (任福民), Weijun Zhu<sup>3</sup> (朱伟军)

<sup>1</sup> Shanghai Typhoon Institute of China Meteorological Administration, Shanghai 200030, China

<sup>2</sup> State Key Laboratory of severe weather, Chinese Academy of Meteorological Sciences, Beijing 100081, China

<sup>3</sup> Key Laboratory of Meteorological Disaster of Ministry of Education, Nanjing University of Information Science & Technology, Nanjing 210044, China

**Abstract** In this paper, typhoon simply means tropical cyclone. As risk is future probability of hazard events, when suppose future probability is the same as historical probability for a specific period, we can understand risk by learning from past events. Based on precipitation and wind data over the mainland of China during 1980 - 2014 and disaster and social data at the county level in Zhejiang province from 2004 to 2012, a study on risk zoning of typhoon disasters (typhoon disasters in this paper refer to affected population or direct economic losses caused by typhoons in Zhejiang province) is carried out. Firstly, characteristics of typhoon disasters and factors causing typhoon disasters are analyzed. Secondly, an intensity index of factors causing typhoon disasters and a population vulnerability index are developed. Thirdly, combining the two indexes, a comprehensive risk index for typhoon disasters is obtained and used to zone areas of risk. The above analyses show that, southeastern Zhejiang is the area most affected by typhoon disasters. The annual probability of the occurrence of typhoon rainstorms >50 mm decreases from the southeast coast to inland areas, with a maximum in the boundary region between Fujian and Zhejiang, which has the highest risk of rainstorms. Southeastern Zhejiang and the boundary region between Zhejiang and Fujian province and the Hangzhou Bay area are most frequently affected by typhoon extreme winds and have the highest risk of wind damage. The population of southwestern Zhejiang is the most vulnerable to typhoons as a result of the relatively undeveloped economy, mountainous terrain and the high risk of geological disasters in this region. Vulnerability is lower in the cities due to better disaster prevention and reduction strategies and a more highly educated population. The southeast coastal areas face the highest risk of typhoon disasters, especially in the boundary region between Taizhou and Wenzhou cities. Although the inland mountainous areas are not directly affected by typhoons, they are in the

---

Corresponding author: Dr. Fumin Ren, State Key Laboratory of severe weather (LaSW)/CAMS, Beijing, 100081.  
E-mail address: fmren@163.com

30 medium-risk category for vulnerability.

31 **Keywords:** typhoon disasters, factors causing typhoon disasters, vulnerability, comprehensive risk

32 index, risk zoning

### 33 **1 Introduction**

34 Typhoon, which means tropical cyclone in this paper, often causes some of the most serious natural  
35 disasters in China, with an average annual direct economic loss of about \$9 billion. The arrival of  
36 typhoon is often accompanied by heavy rain, high winds and storm surges, with the main impacts in  
37 southern coastal areas of China (Zhang et al., 2009). Zhejiang province is seriously affected by  
38 typhoons—for example, in 2006, super-typhoon Sang Mei caused 153 deaths in Cangnan county of  
39 Wenzhou city, with 11.25 billion yuan of direct economic losses. Therefore it would be of practical  
40 significance to develop a system for the risk assessment of typhoon disasters in Zhejiang province.

41 Major risk assessment models include the disaster risk index system of the United Nations  
42 Development Program (global scale, focusing on human vulnerability), the European multiple risk  
43 assessment (with emphasis on factors causing disasters and vulnerability) and the American  
44 HAZUS-MH hurricane module and disaster risk management system. Vickery et al. (2009) and Fang et  
45 al. (2012, 2013) reviewed the factors causing typhoon disasters. Rain and wind are direct causes of  
46 typhoon disasters (Emanuel, 1988, 1992, 1995; Holland, 1997; Kunreuther and Roth, 1998); stronger  
47 typhoons produce heavier rain and stronger winds, resulting in a greater number of casualties and  
48 higher economic losses. Many of the researches on the factors causing typhoon disasters used a grade  
49 index and the probability of occurrence (Chen et al., 2011; Su et al., 2008; Ding et al., 2002; Chen,  
50 2007). Recently, some research built quantitative assessment in some provinces and carried out  
51 preliminary studies on pre-evaluating typhoon disasters (Huang and Wang, 2015; Yin and Li, 2017).

52 In terms of vulnerability, Pielke et al. (1998, 2008) combined the characteristics of typhoons and  
53 socioeconomic factors, suggesting that both the vulnerability of the population and economic factors  
54 were important in estimating disaster losses. The vulnerability of a population is a pre-existing  
55 condition that influences its ability to face typhoon disasters. Among the most widely used indexes is  
56 the Social Vulnerability Index (SoVI) (Cutter et al., 2003; Chen et al., 2011). Other researches have  
57 focused on the vulnerability of buildings, obtaining a fragility curve by combining historical loss with

58 the characteristics of buildings and typhoons (Hendrick and Friedman, 1966; Howard et al., 1972;  
59 Friedman, 1984; Kafali and Jain, 2009; Pita et al., 2014). Studies in China have assessed vulnerabilities  
60 to typhoon disasters (Yin et al., 2010; Niu et al., 2011). Evaluation indexes for the assessment of  
61 disaster losses were established based on the number of deaths, direct economic losses, the area of  
62 crops affected and the number of collapsed houses. These indexes were used to construct different  
63 disaster assessment models (Liang and Fan, 1999; Lei et al., 2009; Wang et al., 2010). Xu et al. (2015)  
64 comprehensively assessed the impact of typhoons across China using the geographical information  
65 system. The future direction of tropical cyclone risk management is quantitative risk models (Chen et  
66 al., 2017).

67 Previous studies have concentrated on semi-quantitative, large-scale research, with less emphasis  
68 on quantitative research at county level based on large amounts of accurate data. In addition, the studies  
69 have paid more attention to disaster losses. Few studies have focused on a comprehensive risk  
70 assessment of typhoon disasters coupled with factors causing typhoon disasters and population  
71 vulnerability. In this study, Zhejiang province, which is frequently affected by the strongest landfall  
72 typhoons (Ren et al., 2008) and experiences most serious typhoon disasters (Liu and Gu, 2002) in the  
73 mainland of China, is selected as the study area. This paper does not consider the impact of storm  
74 surges. The factors causing typhoon disasters are represented by typhoon rain and typhoon wind.  
75 Section 2 introduces the data and methods used in this study. Section 3 provides analyses on typhoon  
76 disaster losses and causing factors. Section 4 presents risk assessment and regionalization of typhoon  
77 disasters. Summary and discussions are given in the final section.

## 78 **2 Data and Methods**

79 This study is carried out in Zhejiang province (Figure 1) including 11 cities along the Yangtze River  
80 Delta. Zhejiang province is in the eastern part of the East China Sea and north to Fujian province,  
81 which is one of the most economically powerful provinces in China.

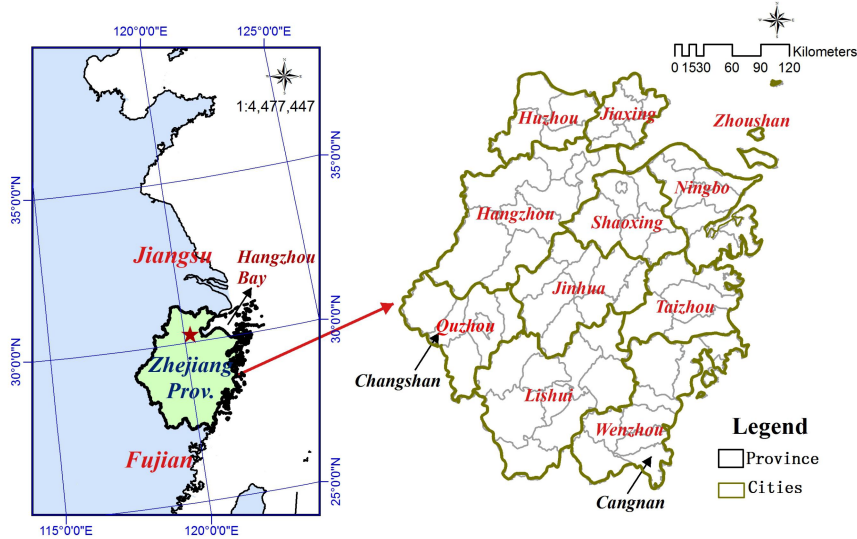


Figure 1. Maps of Zhejiang province, China showing location and major cities.

## 2.1 Data

### 2.1.1 Typhoon, Precipitation and Wind Data

The typhoon data used in this study are the best-track tropical cyclone datasets from Shanghai Typhoon Institute for the time period 1960 - 2014 (Eunjeong and Ying, 2009; Li and Hong, 2015). Daily precipitation data for 2479 stations and daily wind data for 2419 stations during the time period 1960 - 2014 over the mainland of China are obtained from the National Meteorological Information Center. The maximum wind speed is given as the maximum of 10-minute mean. In this paper, two time periods of precipitation and wind data are used.

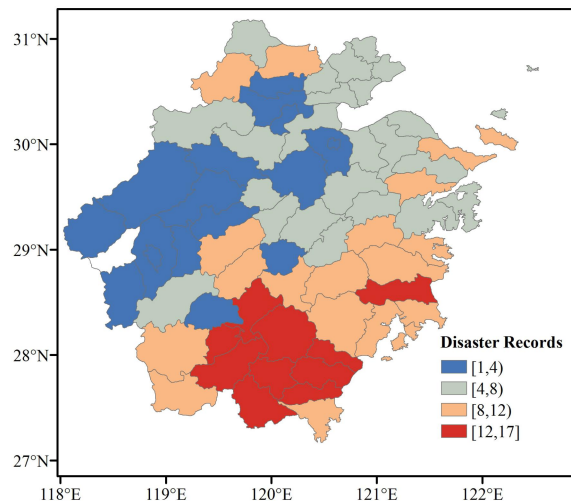
Because of limited access to county-level typhoon disaster data, we have only obtained data during 2004 to 2012. So when calculating intensity index of factors causing typhoon disasters, the time period of typhoon precipitation and typhoon wind are the same as typhoon disasters, which is 2004 - 2012.

For risk analyses of typhoon precipitation and typhoon wind (please see detail in sections 3.1 and 3.2), suppose future probability is the same as historical probability, we then select the period of 1980 - 2014. As Lu et al. (2016) mentioned, considering the homogeneity of wind data, we use the period of 1980 - 2014 for wind analysis. To ensure the consistency between wind and precipitation data, 1980 - 2014 is selected as the period. In addition, the Objective Synoptic Analysis Technique (OSAT) method need to identify typhoon wind and precipitation from a wider range than Zhejiang province (please see

102 details in section 2.2.1), so 2419 stations of precipitation data and 2479 stations of wind data over the  
103 mainland of China are used, which all contained 71 stations corresponding to counties in Zhejiang  
104 province.

### 105 2.1.2 Disaster and Social Data

106 Disaster data for each typhoon that affected Zhejiang province from 2004 to 2012 are obtained from the  
107 National Climate Center and the number of records for each county is shown in Figure 2. Of the 11  
108 cities in Zhejiang province, Wenzhou and Taizhou record the most typhoon disasters, with a maximum  
109 being 17 at Wenzhou. Fewer typhoon disasters are recorded in the central and western regions of  
110 Zhejiang province, particularly in Changshan and Quzhou, which may be because the strength of  
111 typhoons weakened after landfall. The population data in 2010 are obtained from the sixth national  
112 population census (Population Census Office of the National Bureau of Statistics of China), and the  
113 2010 statistical yearbooks of each city in Zhejiang province published by the cities' statistical bureaus.  
114 The census data is updated every six years, and the 2010 census results are exactly during 2004-2012  
115 which is the research period. Therefore, the population data for 2010 in this paper can basically  
116 represent the population vulnerability of this period. Basic geographical data are obtained from the  
117 National Geomatics Center of China.



118  
119 Figure 2. Number of records of typhoon disasters in Zhejiang province from 2004 to 2012.

## 120 2.2 Methods

### 121 2.2.1 Objective Synoptic Analysis Technique

122 The widely used objective synoptic analysis technique (OSAT) proposed by Ren et al. (2001, 2007,  
123 2011) is used to identify precipitation due to typhoons in this study. The OSAT method is a numerical  
124 technique to separate tropical cyclone induced precipitation from adjacent precipitation areas. Based on  
125 structural analysis of precipitation field, it can be divided into different rain belts. Then, according to  
126 the distances between a TC center and these rain belts, typhoon center and each station, typhoon  
127 precipitation is distinguished. Lu et al. (2016) improved the OSAT method and applied it to identify  
128 typhoon winds. With the application of the OSAT method, daily precipitation and wind data over the  
129 mainland of China during 1980 to 2014 are used for identifying typhoon precipitation and wind data.

### 130 **2.2.2 Canonical Correlation Analysis (CCA)**

131 We use the canonical correlation analysis method to determine the relationship between the affected  
132 population, the rate of economic damage, and typhoon precipitation and winds. In statistics, canonical  
133 correlation analysis (CCA) is a way of inferring information from cross-covariance matrices. If we  
134 have two vectors  $X = (X_1, \dots, X_n)$  and  $Y = (Y_1, \dots, Y_m)$  of random variables, and there are correlations  
135 among the variables, then CCA can find linear combinations of the  $X_i$  and  $Y_j$  which have maximum  
136 correlation with each other (Hardoon et al., 2014). The method was first introduced by Hotelling in  
137 1936 (Hotelling, 1936). The main point of CCA is to separate linear combinations of new variables  
138 from the two sets of variables. In this case, the correlation coefficient between new variables reaches  
139 the maximum. In this paper, we chose factors causing typhoon disasters as a set of variables, and  
140 typhoon disaster as another. Under the maximum canonical correlation coefficient, the linear  
141 combination coefficients (typical variable coefficients) of factors causing typhoon disasters can be used  
142 as weight coefficients of this group of variables. Then we can determine the impact of factors causing  
143 typhoon disasters.

### 144 **2.2.3 Data Standardization**

145 We adopt two methods: Z-score standardization and MIN-MAX standardization. The Z-score  
146 standardized method is based on the mean and standard deviation of the raw data, which is prepared for  
147 CCA method. The MIN-MAX standardization is a linear transformation of the original data so that the  
148 original value maps the interval  $[0, 1]$ . Z-score standardization is used for calculating the intensity  
149 index of factors causing typhoon disasters. Both typhoon precipitation and typhoon maximum wind  
150 speed are standardized by this method. When calculating the typhoon disaster comprehensive risk

151 index (R), we use MIN-MAX standardization to standardize the intensity index of the factors causing  
152 typhoon disasters (I) and the population vulnerability index (SoVI).

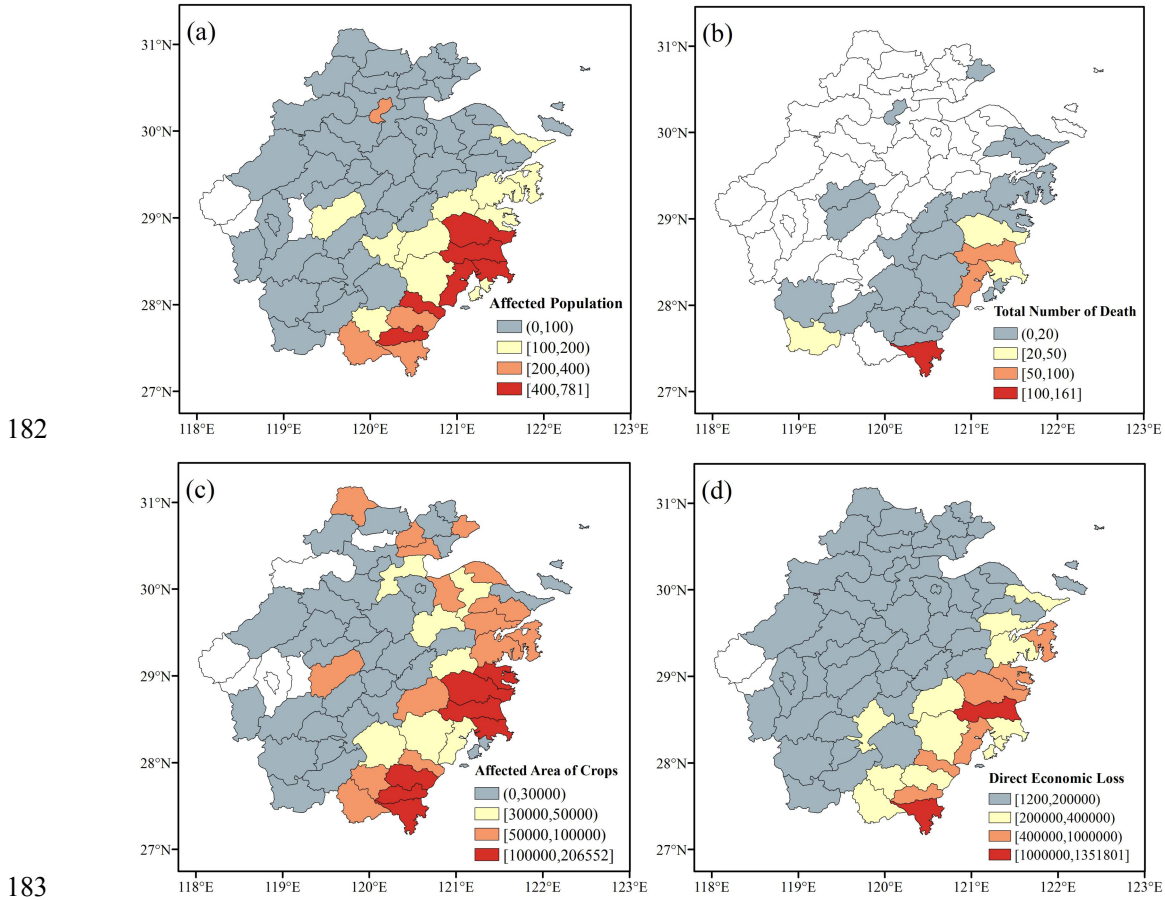
#### 153 **2.2.4 Vulnerability Assessment (SoVI, PCA)**

154 County-level socioeconomic and demographic data are used to construct an index of social  
155 vulnerability to environmental hazards named the Social Vulnerability Index (SoVI). Principal  
156 Component Analysis (PCA) is the primary statistical technique for constructing the SoVI. The PCA  
157 method captures multi-dimensionality by transforming the raw dataset to a new set of independent  
158 variables. Then a few components can represent the dimensional data, and underlying factors can be  
159 identified easily. These new factors are placed in an additive model to compute a summary  
160 score—SoVI (Cutter et al., 2003). Based on various SoVIs derived for disaster social vulnerability in  
161 America, Chen et al. (2014) collects 29 variables as proxies to build a set of vulnerability indexes for  
162 the social and economic environment in China. We then use these vulnerability indexes to calculate the  
163 population vulnerability index for Zhejiang province.

### 164 **3 Typhoon Disaster Losses and Factors**

165 Based on the distribution of typhoon disaster losses in Zhejiang province from 2004 to 2012 (Figure 3),  
166 the affected areas are mainly locates in the southeast corner of the province. The centers with the  
167 largest affected population (Fig. 3a), the largest area of affected crops (Fig. 3c) and the highest direct  
168 economic losses (Fig. 3d) are in Wenzhou and Taizhou cities, although the losses in Ningbo City are  
169 also relatively high. Cangnan in Wenzhou City is the most severely affected, with the highest  
170 cumulative death toll (Fig. 3b). According to the statistical yearbooks of each city in Zhejiang province,  
171 Jiaxing, Shaoxing, Hangzhou in the northeast, and Wenzhou, Jinhua and Taizhou in the southwest are  
172 the regions with the largest agricultural planting area, with more agricultural population in the  
173 southwest. Only parts of the plain area were affected by serious agricultural disasters in the northeast.  
174 The agricultural disaster areas in the southwest are wider. (Fig. 3c). According to the main indicators of  
175 Zhejiang's national economy (total GDP and per capita GDP), the central cities such as Hangzhou in  
176 the northeast had the most developed economy, and the urban economies of Wenzhou and Taizhou in  
177 the southwest were also relatively good. However, the economic losses in southwestern Zhejiang are  
178 severe, much higher than in the northeastern cities. (Fig. 3d). The losses in the affected counties are  
179 associated with the frequency and intensity of typhoons. We therefore analyze the risk of typhoon

180 precipitation and winds in every county in Zhejiang province to provide a reference dataset for the  
 181 factors causing typhoon disasters.



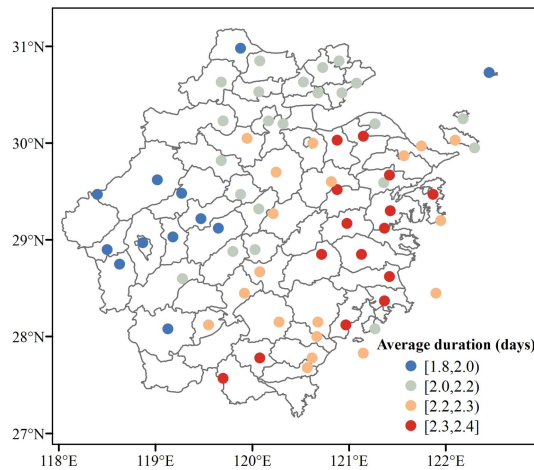
184 Figure 3. Distribution of typhoon disaster losses in Zhejiang province from 2004 to 2012. (a) Affected  
 185 population (millions); (b) total number of deaths (person); (c) area of affected crops (hectares); and (d)  
 186 direct economic losses (millions yuan).

### 187 3.1 Probability of Typhoon Rainstorms

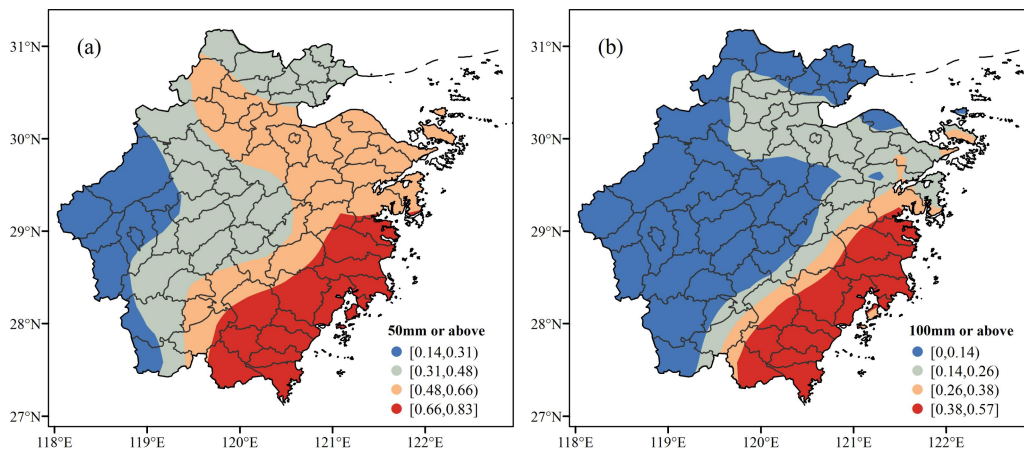
188 The main hazard of typhoon precipitation is concentrated precipitation, so the average duration (days)  
 189 of typhoon precipitation at each station in Zhejiang province is counted from 1980 to 2014 (Figure 4).  
 190 The duration of typhoon rainfall is less in inland areas, especially in Quzhou City. Persistent  
 191 precipitation is concentrated in Wenzhou, Taizhou and Ningbo cities, where there may have been a  
 192 higher risk of typhoon disasters. Typhoon rainstorm in this study means daily typhoon precipitation  
 193 over 50mm, and typhoon torrential rainstorm means daily typhoon precipitation over 100mm. The  
 194 probability is the annual possibility of the occurrence of typhoon rainstorms. The probability  
 195 denominator is the total number of years, and the numerator is the annual frequency of typhoon



196 precipitation. If a station experiences typhoon precipitation in one year, the numerator increases by one.  
 197 Based on the probability of typhoon rainstorms occurring in each county in Zhejiang province (Figure  
 198 5), we found that the annual probability of the occurrence of typhoon rainstorms is highest over the  
 199 southeast coast of Zhejiang province from 1980 to 2014, especially in Taizhou City, where the annual  
 200 probability is 83%. The annual probability of typhoon rainstorms with precipitation >100 mm is lower,  
 201 but the distribution of probability is consistent with the rainstorms with lower precipitation. The  
 202 probability of typhoon torrential rainstorms decreases rapidly in the western and central regions of  
 203 Zhejiang province, although the range increases. There are three centers of high probability: Taizhou,  
 204 Wenzhou and Ningbo cities.



205  
 206 Figure 4. Average duration (days) of typhoon precipitation at each station in Zhejiang province from  
 207 1980 to 2014.

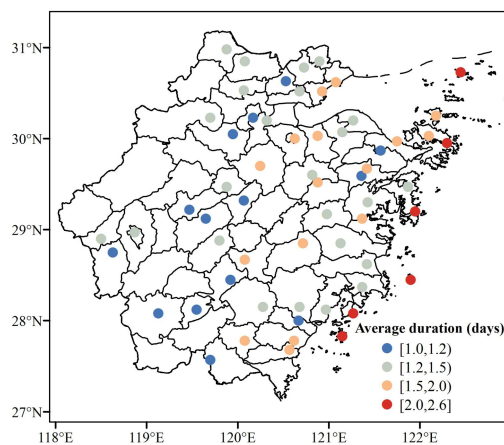


208  
 209 Figure 5. Probability of the occurrence of typhoon rainstorms in Zhejiang province: (a) rainstorms with  
 210 precipitation >50 mm; and (b) torrential rainstorms with precipitation >100 mm.

211 **3.2 Probability of Typhoon Winds**

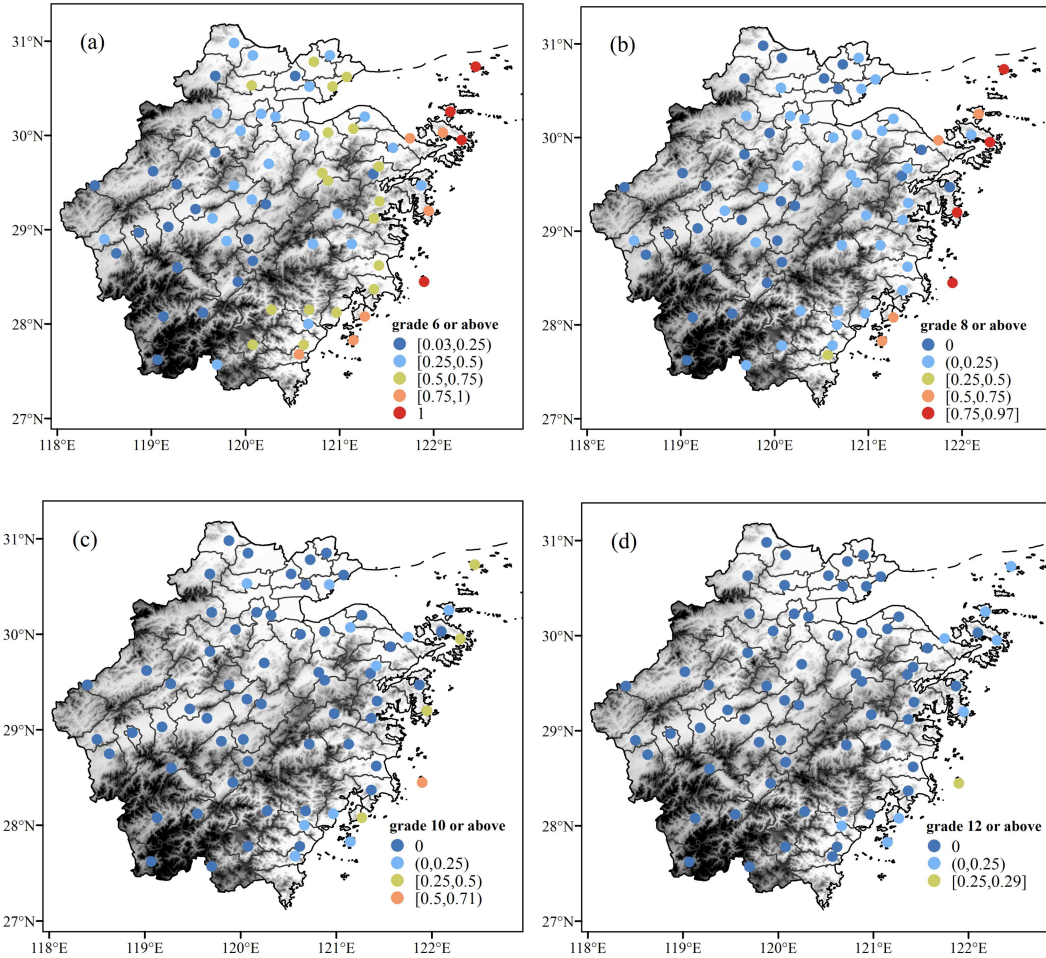
212 The average duration (days) of typhoon winds (over 6 grade) is calculated in Zhejiang province (Figure  
213 6). The duration of typhoon winds is relatively short in the central and western regions and the typhoon  
214 winds are concentrated in the coastal areas of Wenzhou, Taizhou and Ningbo cities. The longest  
215 duration of typhoon winds occurs over the offshore islands.

216 The main hazard from typhoon winds is manifested in the destructive force of strong winds.  
217 Therefore we calculate the probability of annual occurrence of typhoon winds at or above grades 6 and  
218 12 at each station from 1980 to 2014 (Figure 7). Typhoon winds at or above grade 6 mainly occur along  
219 the coastal areas, with rare occurrence in the mountainous areas. Meanwhile, the probability of typhoon  
220 winds at or above grade 8 is generally 0.5~0.9 along the coast, and below 0.25 in the inland  
221 mountainous areas. Typhoons winds at or above grade 10 or 12 are much less likely and are only seen  
222 in the coastal areas and islands, with a rapidly decreasing probability from the coastal areas to the  
223 inland mountainous areas. The areas at high probability of typhoon winds are consistent with those  
224 with a high probability of typhoon rain, i.e. Wenzhou, Taizhou and Ningbo cities. The probability of  
225 typhoon extreme winds is much higher in coastal areas than inland.



226

227 Figure 6. Average duration (days) of typhoon winds (over 6 grade) at each station in Zhejiang province  
228 from 1980 to 2014.



229

230

231 Figure 7. Probability of the occurrence of typhoon winds in Zhejiang province at (a) grade 6 or above  
 232 ( $\geq 10.8$  m/s), 672 (b) grade 8 or above ( $\geq 24.5$  m/s), (c) grade 10 or above ( $\geq 32.7$  m/s) and (d) grade 12  
 233 or above ( $\geq 41.5$  m/s).

234

## 235 4 Risk Assessment and Regionalization of Typhoon Disasters

### 236 4.1 Intensity Index of Factors Causing Typhoon Disasters

237 The main factors causing typhoon disasters, which are considered in this study, are rainstorms and  
 238 winds. The level and intensity of a single factor cannot fully represent and describe the impact. It is  
 239 necessary to determine their influence through typical correlation analysis, and then typhoon wind and  
 240 rain effect are superimposed by the weight coefficients. Therefore we establish a comprehensive  
 241 intensity index that includes typhoon precipitation and winds. Taking the county as a unit, we select all  
 242 the typhoons that affected the population of Zhejiang province from 2004 to 2012. The total

243 precipitation and daily maximum wind speed during typhoons measured in each county are used to  
 244 describe the factors causing typhoon disasters. The total sample size is 322. Using CCA, we determine  
 245 the impact of typhoon precipitation and winds on the population. We then do CCA for all the typhoons  
 246 that caused direct economic losses in Zhejiang province from 2004 to 2012, and the total sample size is  
 247 404 (Table 1). The effect of typhoon precipitation on both the population and direct economic losses is  
 248 always greater than that of typhoon winds. By averaging typical coefficients for both precipitation and  
 249 wind, weight coefficients of 0.85 and 0.65 are obtained within the intensity index for precipitation and  
 250 winds, respectively.

251 Table 1. Canonical correlation analysis of factors causing typhoon disasters.

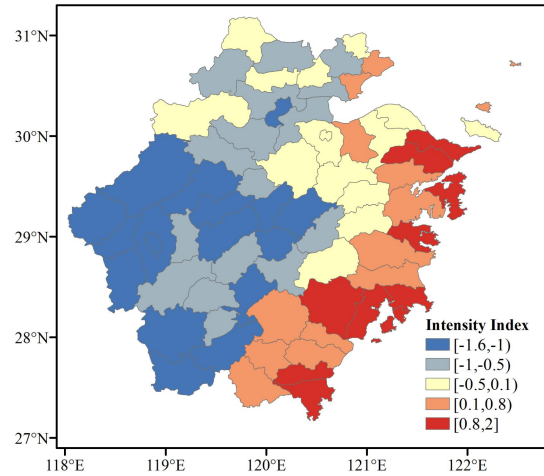
Disasters	Canonical correlation coefficient	Canonical variable coefficient	
		Typhoon precipitation	Typhoon wind
Affected population	0.45	0.84	0.651
Direct economic losses	0.477	0.863	0.655

252

253 Based on the weight coefficients in Table 1, an intensity index of factors causing typhoon  
 254 disasters is established:

$$255 \quad I = Ax + By \quad (1)$$

256 where  $I$  is the intensity index of factors causing typhoon disasters,  $X$  is the standard typhoon  
 257 precipitation and  $Y$  is the maximum wind speed of the typhoon.  $A$  and  $B$  are the weight coefficients for  
 258 typhoon precipitation and typhoon winds, respectively. Using Equation (1), we average the intensity  
 259 indexes of typhoons at each station (Figure 8). Based on the distribution of these average intensity  
 260 indexes, three high value centers, namely Wenzhou, Taizhou and Ningbo cities are identified, which is  
 261 consistent with the results of Chen et al. (2011), can be found.



262

263 Figure 8. Intensity indexes of factors causing typhoon disasters at each station in Zhejiang province.

264 **4.2 Population Vulnerability Index**

265 Natural disasters are social constructions and the basic causes of losses are the attributes of human  
 266 beings and their social system (Jiang 2014). The index system of Chen et al. (2011) is used to evaluate  
 267 the vulnerability of Zhejiang province. Based on the extracted population information, 29 variables are  
 268 identified that may affect vulnerability (Table 2).

269

Table 2. The 29 variables affecting vulnerability in Zhejiang province.

	variables	Name
	Per capita disposable income of urban residents	
1	(yuan)	UBINCM
2	Percentage of female (%)	QFEMALE
3	Percentage of minority (%)	QMINOR
4	Median age	MEDAGE
	Unemployment rate (calculated - unemployed	
5	population / (unemployed + total population)	QUNEMP
6	Population density	POPDEN
7	Percentage of urban population (%)	QUBRES
	Percentage of non-agricultural household	
8	population (%)	QNONAGRI
	Percentage of households that living in rented	
9	houses (%)	QRENT
	Percentage of employees working in primary	
10	industries and mining (%)	QAGREMP
	Percentage of employees working in secondary	
11	industries (%)	QMANFEMP
	Percentage of employees working in tertiary	
12	industries (%)	QSEVEMP

13	Household size (person / household)	PPUNIT
14	Percentage of population with college degree (25 years old and older)	QCOLLEGE
15	Percentage of population with high school degree (20 years old and older)	QHISCH
16	Percentage of illiterate people (15 years old and older)	QILLIT
17	Population growth rate (2000-2010)	POPCH
18	Average number of rooms per household (inter / household)	PHROOM
19	Per capita housing construction area (m <sup>2</sup> / person)	PPHAREA
20	Percentage of premises without tap water (%)	QNOPIPWT
21	Percentage of premises without a kitchen (%)	QNOKITCH
22	Percentage of premises without a toilet (%)	QNOTOILET
23	Percentage of premises without a bath (%)	QNOBATH
24	Number of beds per 1000 person in health care institutions	HPBED
25	Number of medical personnel per 1000 resident population	MEDPROF
26	Percentage of people under 5	QPOPUD5
27	Percentage of population over 65 years old	QPOPAB65
28	Population dependency ratio (%)	QDEPEND
29	Percentage of population covered by subsistence allowances (%)	QSUBSIST

270 After Principal Component Analysis (PCA) of the 29 variables, seven components with  
271 eigenvalue >1 are extracted. Based on the variable meanings in each component, these 7 components  
272 are named as table 3. The first component, which reflects the income of the population and the  
273 employment situation, contribute 30.1% of the total variance. This component is positive because the  
274 more property there is in an area, the higher the vulnerability to damage. The second component, which  
275 reflects education level of the population, occupies 15.6% of the total variance. This component is  
276 negative because if education level is higher, then the population's awareness of disaster prevention and  
277 reduction is greater and their vulnerability is lower. The third component, which reflects the number of  
278 dilapidated houses, takes up 8.7% of the total variance. This component plays a positive part in  
279 vulnerability. The fourth component, which reflects the illiteracy and the number of young people, is  
280 positive and represents 8.4% of the total variance. The fifth component, which reflects the household  
281 size and the percentage of women, explains 7.7% of the total variance and is positive. The sixth  
282 component, which reflects the number of ethnic minorities, contributes 6.1% of the total variance and  
283 is positive. The seventh component, which represents 5.3% of the total variance, reflects the

284 unemployment rate and the housing area and is positive.

285 The total variance explained by these seven components is up to 81.9%, which can be used to  
 286 represent the population vulnerability of Zhejiang province. The distributions of the first (positive)  
 287 component and the second (negative) component are shown in Figure 9. Areas with a low employment  
 288 rate have high vulnerability, but the vulnerability is low in urban areas with higher levels of education.  
 289 The seven components thus represent the real situation of the population vulnerability in Zhejiang  
 290 province to the effect of typhoons. The population vulnerability index in Zhejiang province (SoVI) is  
 291 calculated as:

$$292 \quad \text{SoVI} = \text{component 1} - \text{component 2} + \text{component 3} + \text{component 4} + \text{component 5} + \text{component} \\ 293 \quad 6 + \text{component 7} \quad (2)$$

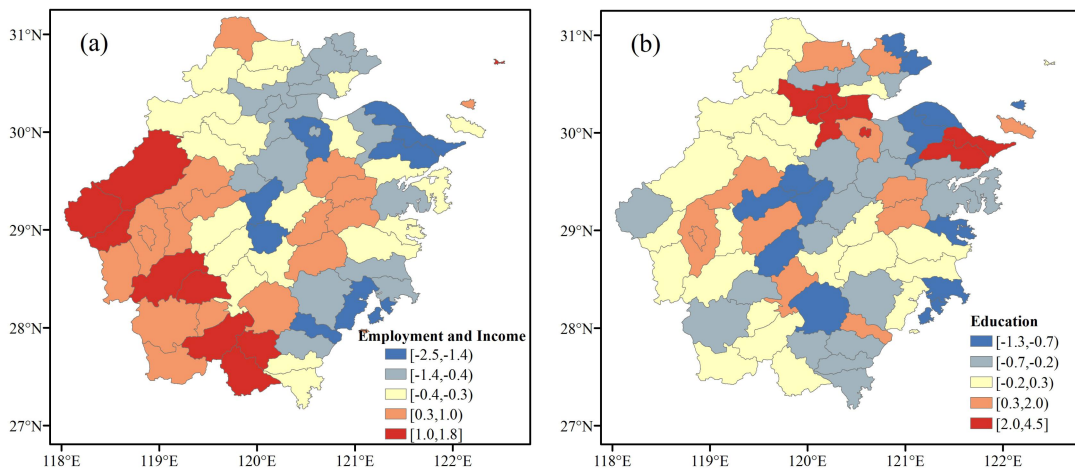
294 By calculating the vulnerability indexes of each county, the distribution of population  
 295 vulnerability in Zhejiang province is obtained (Figure 10). The areas with high vulnerabilities are  
 296 mountainous regions where the economy is relatively undeveloped, whereas the vulnerability is low in  
 297 cities, such as Hangzhou and Huzhou cities, where there is a greater awareness of disaster prevention  
 298 and reduction and houses are of high quality.

299 Table 3. The seven components extracted by PCA.

Components	Contained variables	Name	(Sign)
1	QMANFEMP, UBINCM, QAGREMP, QRENT, POPCH, QDEPEND, QSUBSIST, QPOPAB65, POPDEN, MEDAGE, QNOKITCH, QILLIT, PHROOM, PPHAREA	Employment and poverty	(+)
2	QHISCH, QCOLLEGE, QNONAGRI, QSEVEMP, HPBED, MEDTECH	Education	(-)
3	QNOBATH, QNOTOILET, PPUNIT	Number of dilapidated houses	(+)

4	QILLIT, QDEPEND, QPOPUD5, MEDAGE	Illiteracy and juvenile population	(+)
5	QFEMALE, PHROOM, PPHAREA, QSEVEMP	Household size and ratio of women	(+)
6	QMINOR	Ethnic minority	(+)
7	QUNEMP, QNOPIPWT	Unemployment and housing size	(+)

300



301

302 Figure 9. Distribution of population vulnerability index of (a) component 1 (employment and

303 income) and (b) component 2 (education).



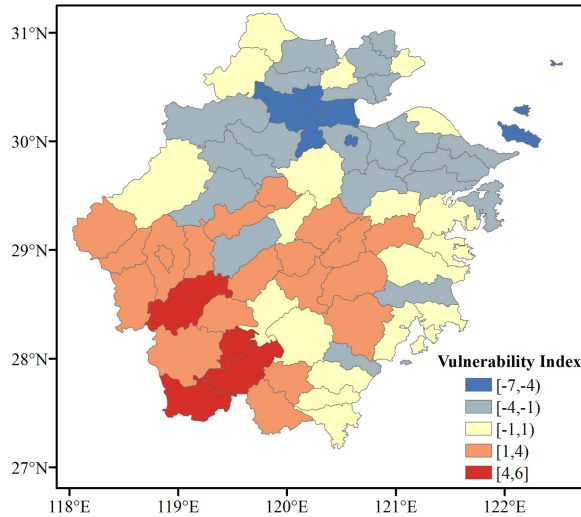


Figure 10. Distribution of population vulnerability index of counties.

### 4.3 Typhoon Disaster Comprehensive Risk Index and Zoning

The typhoon disaster risk assessment system is mainly composed of the factors causing disasters, the population vulnerability and the environment. In this paper, typhoon disaster comprehensive risk index is obtained by combining the factors causing typhoon disasters and vulnerability, without taking the sensitivity of the environment into account. After standardizing the intensity index of factors causing typhoon disasters and the population vulnerability index, the typhoon disaster comprehensive risk index ( $R$ ) is obtained as follows:

$$R = \text{intensity index of factors causing typhoon disasters } (I) \times \text{vulnerability index (SoVI)} \quad (3)$$

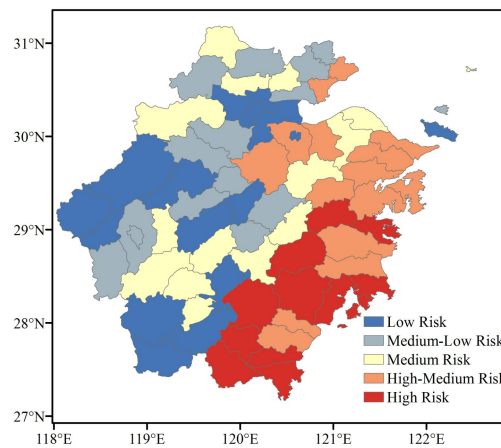
Based on the comprehensive risk index, five risk grades for typhoon disasters are defined (Table 4), and risk zoning of typhoon disasters in Zhejiang province has been done as shown in Figure 11. The classification of typhoon disaster risk index is based on the natural breaks method (Jenks) provided by Arcgis.

Table 4. Disaster risk index and grading.

Risk grade:	High	High–medium	Medium	Medium–low	Low
Risk index:	0.3	0.18–0.3	0.13–0.18	0.07–0.13	0.07

Figure 11 shows that, the index presents a good reflection of the distribution of typhoon disasters in Zhejiang province (Figure 3), especially in the southeastern coastal areas. The southeast coastal areas face the highest risk, especially in the boundary regions between Zhejiang and Fujian province, and

322 Taizhou and Wenzhou cities. Overall, the risk of typhoon disasters decreases from the coast to inland  
323 areas. Cities are at medium to low risk as a result of their developed economy, high-quality houses and  
324 better educated population. The inland mountainous areas have a high vulnerability. Although they are  
325 not directly affected by typhoons, they are still in the middle risk areas as a result of their poorly  
326 developed economy.



327

328 Fig. 11. Risk zoning of typhoon disasters in Zhejiang province.

## 329 5 Discussion and Conclusions

330 (1) An intensity index of factors causing typhoon disasters is developed, with highest values in  
331 Wenzhou, Taizhou and Ningbo cities. A comparison between the distributions of the intensity index and  
332 actual typhoon disasters in Zhejiang province from 2004 to 2012 shows that the index is a good  
333 reflection of the possibility of typhoon disasters.

334 (2) Seven components are extracted after PCA of 29 variables affecting vulnerability. These seven  
335 factors represent 81.9% of the total variance and are a good reflection of the index of population  
336 vulnerability in Zhejiang province. Southwestern Zhejiang is the most vulnerable as it has a relatively  
337 undeveloped economy, more mountainous areas and a higher risk of geological disasters.  
338 Vulnerabilities are lower in cities as a result of better disaster prevention and reduction measures and a  
339 better educated population.

340 (3) Typhoon disaster comprehensive risk index is obtained by combining the factors causing  
341 typhoon disasters and population vulnerability. Based on the comprehensive risk index, risk zoning of  
342 typhoon disasters in Zhejiang province is achieved. The southeast coastal areas are at high risk,  
343 especially the boundary regions between Zhejiang and Fujian province, and Taizhou and Wenzhou

344 cities. The risk of typhoon disasters decreases quickly from coastal areas to inland regions. Cities are at  
345 medium to low risk because of their developed economy, high-quality houses and better educated  
346 population.

347 Although some interesting results have been obtained in this study, there are still some problems  
348 that require further study. As a result of the limited data on typhoon disasters, it is currently impossible  
349 to give a long time trend for high-resolution typhoon disaster analysis. It is also unclear whether this  
350 methodology can be applied to other regions. This paper mainly considers the effects of typhoon rain  
351 and typhoon wind, without considering the impact of storm surge. This is the limitation of the study,  
352 and we will explore the role of storm surges in future work.

### 353 **Acknowledgments**

354 This study is supported by the Chinese Ministry of Science and Technology Project No.  
355 2015CB452806.

### 356 **References**

- 357 Chen, H. Y., Yan, L. N., and Lou, W. P.: On assessment indexes of the strength of comprehensive  
358 impacts of tropical cyclone disaster-causing factors, *Journal of Tropical Meteorology*, 27(1), 139-144,  
359 2011. (in Chinese)
- 360 Chen, W. F., Xu, W., and Shi, P. J.: Risk assessment of typhoon disaster at county level in the Yangtze  
361 river delta of China, *Journal of Natural Disasters*, 4, 77-83, 2011.
- 362 Chen, W. F., Duan, Y. H., and Lu, Y.: Review on Tropical Cyclone Risk Assessment, *Journal of*  
363 *Catastrophology*, 32(4), 2017. (in Chinese)
- 364 Chen, X.: Vulnerability diagnosis and assessment of typhoon disaster system at coastal regions, a case  
365 study of Fujian province, *Journal of Catastrophology*, 22(3), 6-10, 2007. (in Chinese)
- 366 Cutter, S. L., Boruff, B. J., and Shirley, W. L.: Social Vulnerability to Environmental Hazards, *Social*  
367 *Science Quarterly*, 84(2):242-261, 2003.
- 368 Ding, Y. and Shi, P. J.: Fuzzy risk assessment model of typhoon hazard. *Journal of Natural*  
369 *Disasters*, 11(1), 34-43, 2002. (in Chinese)
- 370 Emanuel, K. A.: The maximum intensity of hurricanes, *Journal of the Atmospheric Sciences*, 45(7),  
371 1143-1155, 1988.
- 372 Emanuel, K. A.: The dependence of hurricane intensity on climate, *American Institute of Physics*, 277,  
373 25-33, 1992.
- 374 Emanuel, K. A.: Sensitivity of tropical cyclones to surface exchange coefficients and a revised  
375 steady-state model incorporating eye dynamics, *Journal of the Atmospheric Sciences*, 52(22),  
376 3969-3976, 1995.

377 Eunjeong, C. and Ying, M.: Comparison of three western North Pacific tropical cyclone best track  
378 datasets in seasonal context, *Journal of the Meteorological Society of Japan*, 89(3):211-224, 2009.

379 Fang, W. H. and Lin, W.: A review on typhoon wind field modeling for disaster risk  
380 assessment, *Progress in Geography*, 32(6), 852-867, 2013. (in Chinese)

381 Fang, W. H. and Shi, X. W.: A review of stochastic modeling of tropical cyclone track and intensity for  
382 disaster risk assessment, *Advances in Earth Science*, 27(8), 866-875, 2012. (in Chinese)

383 Friedman, D. G.: Natural hazard risk assessment for an insurance program, *The Geneva Papers on Risk  
384 and Insurance - Issues and Practice*, 9(1), 57-128, 1984.

385 Hardoon, D. R., Szedmak, S., and Shawetaylor, J.: Canonical Correlation Analysis: An Overview with  
386 Application to Learning Methods, *Neural Computation*, 16(12):2639-2664, 2014.

387 Hendrick, R. L. and Friedman, D. G.: Potential impacts of storm modification on the insurance industry,  
388 University of Chicago Press, Chicago, 227-248, 1966.

389 Holland, G. J.: The maximum potential intensity of tropical cyclones, *Journal of Atmospheric  
390 Sciences*, 54(21), 2519-2541, 1997.

391 Hotelling, H.: Relations between two sets of variates, *Biometrika*, 28(3/4), 321-377, 1936.

392 Howard, R. A., Matheson, J. E., and North, D. W.: The decision to seed hurricanes. *Science*, 176(4040),  
393 1191-202, 1972.

394 Huang, W. K. and Wang, J. J.: Typhoon damage assessment model and analysis in Taiwan, *Natural  
395 Hazards*, 79(1), 497-510, 2015.

396 IPCC: *Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation*,  
397 Cambridge, UK, and New York, NY, USA, 2012.

398 Jiang, T., Li, X. C., and Chao, Q. C.: Highlights and understanding of climate change 2014: impacts,  
399 adaptation, and vulnerability, *Advances in Climate Change Research*, 10(3), 157-166, 2014.

400 Kafali, C. and Jain, V.: A Methodology for Estimating Typhoon Losses to Building Environment in  
401 Japan, in: *Proceedings of the 11th Americas Conference on Wind Engineering*, San Juan, Puerto Rico,  
402 2009.

403 Kunreuther, H. and Roth, R. J.: *The Status and Role of Insurance Against Natural Disaster in the  
404 United States*, Washington Joseph Henry Press, 1998.

405 Lei, X. T., Chen, P. Y., Yang, Y. H., and Qian, Y. Z.: Characters and objective assessment of disasters  
406 caused by typhoons in china, *Acta Meteorologica Sinica*, 67(5), 875-883, 2009. (in Chinese)

407 Li, S. H. and Hong, H. P.: Use of historical best track data to estimate typhoon wind hazard at selected  
408 sites in China, *Natural Hazards*, 76(2):1395-1414, 2015.

409 Liang, B. Q., Liang, J. P., and Wen, Z. P.: Study of typhoon disasters and its affects in china, *Journal of  
410 Natural Disasters*, 1, 84-91, 1995. (in Chinese)

411 Liang, B. Q. and Fan, Q.: A fuzzy mathematic of the disaster bytropical cyclones, *Journal of Tropical  
412 Meteorology*, 4, 305-311, 1999. (in Chinese)

413 Liu, T. J. and Gu, J. Q.: A statistical analysis of typhoon disasters in Zhejiang province, *Journal of*

414 Catastrophology, 17 (4): 64-71, 2002. (in Chinese)

415 Lu, Y., Zhu, W. J., Ren, F. M., and Wang, X.: Changes of Tropical Cyclone High Winds and Extreme  
416 Winds During 1980-2014 over China, *Advances in Climate Change Research*, 12(5), 413-421, 2016.  
417 (in Chinese)

418 Niu, H. Y., Liu, M., and Lu, M.: Risk assessment of typhoon disasters in china coastal area during last  
419 20 years, *Scientia Geographica Sinica*, 31(6), 764-768, 2011. (in Chinese)

420 Pielke, R. A. J. and Landsea, C. W.: Normalized hurricane damages in the United States:  
421 1925-95, *Weather & Forecasting*, 13(3), 621--631, 1998.

422 Pielke, R. A. J., Gratz, J., and Landsea, C. W.: Normalized hurricane damage in the United States:  
423 1900–2005, *Natural Hazards Review*, 9(1), 29-42, 2008.

424 Pita, G., Pinelli, J. P., and Gurley, K.: State of the art of hurricane vulnerability estimation methods: a  
425 review, *Natural Hazards Review*, 16(2), 04014022, 2014.

426 Ren, F. M., Byron, G., and David, E.: A Numerical Technique for Partitioning Cyclone Tropical  
427 Precipitation, *Journal of Tropical Meteorology*, 17(3), 308-313, 2001. (in Chinese)

428 Ren, F. M., Wang, Y., Wang, X., and Li, W.: Estimating Tropical Cyclone Precipitation from Station  
429 Observations, *Advances in Atmospheric Sciences*, 24(4), 700-711, 2007.

430 Ren, F. M., Wang, X. L., Chen, L. S., and Wang, Y. m.: Tropical cyclones landfalling on mainland  
431 China, Hainan and Taiwan and their correlations, *Acta Meteorologica Sinica*, 66 (2): 224-235, 2008. (in  
432 Chinese)

433 Ren, F. M. and Wu, G. X.: Tropical cyclone over the past 60 years, China Meteorological Press, Beijing,  
434 2011. (in Chinese)

435 Su, G. L., Miao, C. M., and Mao, Y. D.: Typhoon hazard in zhejiang province and risk assessment of its  
436 influence on agriculture, *Journal of Natural Disasters*, 17(5), 113-119, 2008. (in Chinese)

437 Vickery, P. J., Masters, F. J., and Powell, M. D.: Hurricane hazard modeling: the past, present, and  
438 future, *Journal of Wind Engineering & Industrial Aerodynamics*, 97(7–8), 392-405, 2009.

439 Wang, X. R., Wang, W. G., and Ma, Q. Y.: Model for general grade division of typhoon disasters and  
440 application, *Meteorological Monthly*, 36(1), 66-71, 2010. (in Chinese)

441 WMO: Seventh International Workshop on Tropical Cyclones (IWTC-VII), available at:  
442 <http://www.wmo.int/pages/prog/arep/wwrp/tmr/IWTC-VII.html/>, 2010.

443 Xu, X., Sun, D., and Guo, T.: A systemic analysis of typhoon risk across china, *Natural Hazards*, 77(1),  
444 461-477, 2015.

445 Yin, Y. Z. and Li, H. L.: Preliminary study on pre-evaluation method of typhoon disaster in China,  
446 *Meteorological Monthly*, 43(6):716-723, 2017. (in Chinese)

447 Yin, Z. E., Xu, S. Y., Yin, J., and Wang, J.: Small-scale Based Scenario Modeling and Disaster Risk  
448 Assessment of Urban Rainstorm Water-logging, *Acta Geographica Sinica*, 65(5), 553-562, 2010. (in  
449 Chinese)

450 Zhang, Q., Wu, L. G., and Liu, Q. F.: Tropical cyclone damages in china 1983-2006, *Bulletin of the*  
451 *American Meteorological Society*, 90(4), 489-495, 2009.