



1 Brief communication: Comprehensive Resilience to Typhoon Disasters: 2 An Urban Assessment of 27 Cities in Seven Major River Basin, China

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7 **Abstract.** The urban resilience based on typhoon disasters are often not assessed. In this communication, we reflect on
8 this issue by analyzing 27 cities around seven major river basin in mainland China. In specific, we build a
9 comprehensive indicator-based model, and adopted the entropy-weighting TOPSIS method. Results show that the *Hai*
10 *River Basin* and the provincial capitals had a higher resilience to typhoon than others, while cities of the *Pearl River*
11 *Basin* are weaker. In some regions with weaker economy, however, the resilience was relatively higher partly
12 attributing to infrastructure, water conservation projects, and level of information disclosure. The analysis is helpful
13 for agencies and professionals to enhance urban capability of resilience, and provides a realistic reference in response
14 to typhoon threats.

15

16 1. Introduction

17 Typhoons are among the most damaging natural hazards around the globe. According to the report of U.N
18 Office for Disaster Risk Reduction (UNDRR), natural hazard by disasters exceed \$3000 billion economic loss and
19 4.5 billion death or injured worldwide between 1998 and 2020. Of the recorded disasters, typhoon is one of the
20 frequent hazards, which caused widespread damage in particularly those coastal areas and become a growing
21 threat as global climate change evolves (Pauline, et al., 2020; Zhang, et al., 2020). China has long suffered from
22 frequent floods and rainstorm caused by typhoon, resulting in significant losses and casualties (China's MEM,
23 2019). Accordingly, since the 2010s, the China central administration has been committing to promote a series of
24 actions of disaster reduction to enhance resilience. The strategy of resilience has also been written into China's
25 long-term 14th Five-year Development Plan and the 2035 *Vision Goal Outline*. By the statement of UNDRR, the
26 core annotation of "resilience" is taken as a capacity test in a pressure environment to achieve quick recovery
27 from disasters and enhance strategies of risk mitigation through overall adaptive measures (UNDRR, 2018). For
28 many decades, the disaster resilience in fighting against typhoon received heated discussions in related domains,
29 such as geographic science, geology, and physics. Meanwhile, how to measure disaster resilience in a specific
30 societal scenarios has obtained more attention both at home and abroad (Peng, 2023), with emerging indicators
31 for assessment (Yang, 2022; Sawaid, 2020) as well as analysis by multidisciplinary methods and multi-criteria
32 techniques (Huang and Ling, 2018; Bao, et al., 2019; Ping, et al. 2021).

33 Although researchers have been studying resilience related to the influence of typhoon-induced disasters on
34 specific dimensions or regions (Siebeneck et al., 2015; Moghadas, 2019), systematic mapping of large river basin
35 areas or habitats potentially exposed to typhoon is relatively limited, particularly in China with complex
36 geographical conditions. There remain gaps in reflecting the typhoon heterogeneity in terms of demographics,
37 lifeline, and critical infrastructure. This could also be attributed to the fact that the magnitude of typhoon
38 resilience of regions and potential for cascade effects are difficult to qualitatively and quantitatively estimate.

39 This brief communication aims at bringing up the subject of urban resilience on the societal system by
40 highlighting the impacts reported after several typhoon disasters upon China river basins since the 2010s.
41 Specifically, we take both natural and social dimensions into account and construct a comprehensive indicator-
42 based frame, covering the economic, social, environmental and administrative aspects. By applying the entropy
43 weighted TOPSIS method, we measured the urban resilience of 27 key cities in the seven major river basins in
44 China (the Songhua River, the Liaohe River, the Haihe River, the Yellow River, the Yangtze River, the Pearl River,
45 and the Nandu River), considering the heterogeneity of typhoon-prone regions and the accessibility of data. A
46 greater consciousness on the relevance of the resilience topics is essential to foster additional investigation and
47 attention in other spacial geographical contexts.



48 **2. Assessing indicators and method**

49 **2.1 Indicators**

50 The World Meteorological Organization (WMO) considers a typhoon a tropical cyclone with central sustained
 51 winds of force rating 12 to 13. According to China's Climate Bulletin, China incurred 345.45 billion yuan of direct
 52 economic losses from various types of disasters in 2023, of which typhoons caused a total of 47.49 billion yuan. In
 53 2022, China's National Disaster Reduction Commission (NDRC) issued the Plan for Comprehensive Disaster
 54 Prevention and Mitigation, which stated that it will strengthen all-phrase control, and establish a comprehensive
 55 risk-based database on natural hazards including typhoon. By the report from the State Flood Control and
 56 Drought Relief Headquarters, there are seven major river basins in China's coastal regions that more affected by
 57 typhoon including the *Yangtze, Yellow, Pearl, Hai, Huai, Liao, and Songhua Rivers*, and these river basins cover 27
 58 large and medium-sized key cities threatened by typhoon annually. Compared to inland cities, cities within the
 59 river basin are at a higher immediate risk of being struck by typhoon disaster, imposing stronger demand for
 60 resilience enhancement.

61 After searching literature both in China and abroad, departmental policies, and disaster-based studies, we
 62 identified 19 indicators for resilience assessment on typhoon basis, which are abstracted into four dimensions(see
 63 [Table1](#)). In specific, *the economic dimension* herein measures the overall economic capacity in a city, which
 64 encompasses the financial capacity, individual economic support capacity, and health care capacity. Usually
 65 struck by a typhoon disaster, urban response and post-disaster reconstruction is closely related to its economic
 66 support from the local government and status of individuals. The investment in water conservation and disaster
 67 reduction was more stressed for fiscal inputs. *The social dimension* describes the demographic and medical level in
 68 the city region. The demographic aspect contains two correlated indicators. Densely populated areas are
 69 vulnerable to typhoon disasters due to its production activities, while on the contrary, urban areas with lower
 70 population density are generally less victimized. Consequently, the population density usually serves as a
 71 negative correlation relative to urban resilience for typhoons. The number of hospital and health center beds per
 72 1,000 people is a vital indicator that can reflect medical level in a city to typhoon risks. *The environmental*
 73 *dimension* includes the natural and man-made condition, in which the former includes precipitation positively
 74 linked with typhoon, and the latter incorporates human modification of the natural environment coupled with
 75 urban construction, such as the area sown by crops, the green space coverage rate in built-up areas, the quality
 76 rating of water conservancy projects, and the storage level of reservoirs. *The administrative dimension* herein is to
 77 measure the response capacity to typhoon, containing the level of urban lifeline project construction, the
 78 performance of disaster prevention construction, the development level of water management team construction,
 79 and the level of water sector information service.

80 **Table 1.** Indicators of urban resilience for typhoon disaster

| Dimension | Factor | Indicator | Direction |
|---------------|---|---|-----------|
| Economic | Economic development level | Per capita GDP (yuan) | Positive |
| | financial support capability | Per capita fiscal expenditure (yuan) | Positive |
| | Personal financial support capability | Per capita disposable income of residents (yuan) | Positive |
| | Medical security support capability | Proportion of healthcare fiscal expenditure (%) | Positive |
| Social | Demographic factors | Density of population (person/sq.km) | Negative |
| | | The proportion of urban permanent population aged 65 and above, and 14 and below to the total | Negative |
| | Medical and health conditions | Number of beds in hospitals and health centers per thousand people (unit) | Positive |
| Environmental | Climate conditions | Annual precipitation (mm) | Negative |
| | | Annual average temperature (°C) | Negative |
| | Land factors | Annual surface runoff (100 million cu.m) | Positive |
| | | Crop sowing area (thousand hectares) | Positive |
| Water project | Density of green space in built-up areas (%) | Positive | |
| | Quality evaluation level of water conservancy engineering | Positive | |



| | | | |
|------------|---|---|----------|
| | construction level | (level) | |
| | Lifeline engineering | Density of drainage pipelines (km/sq.km) | Positive |
| | construction level | Per capita area of paved roads (sq.m) | Positive |
| Managerial | Disaster prevention | The number of national comprehensive disaster reduction demonstration communities(unit) | Positive |
| | construction effectiveness | The number of water resources and environmental management professionals per ten thousand population (unit) | Positive |
| | Construction of water affair management team | Total number of online information published by water conservancy bureau (pcs) | Positive |
| | Service level of water information disclosure | | |

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82 **2.2 Method**

83 Concerning the analytical dimensions of resilience proposed by Bruneau (2003) and the DROP model proposed
 84 by Cutter (2014), we delineated urban resilience to typhoon disaster into four dimensions as stipulated in the
 85 section above. This categorization approach intuitively mirrored the economic and social dynamics within urban
 86 settings and bore significant relevance to the managing typhoon disaster.

87 Meanwhile, the TOPSIS method was in specific introduced herein, which is a sorting algorithm tool to
 88 approximates optimal solution. This method exhibits flexibility in accommodating various data distribution
 89 states and sample sizes, boasting an uncomplicated calculation process and a broad applicability range (Pishyar
 90 et al., 2016; Hajek, et al., 2019; Zhang et al., 2021). We briefly instruct the operational steps in this study as follows:

91 First, the indicator weights based on the entropy weight method were used to measure the standardized
 92 judgment matrix:

$$93 \quad v_{ij} = \omega_j x'_{ij} \tag{1}$$

94 Second, the positive ideal solution v_j^+ and the negative ideal solution v_j^- for each selected indicator are
 95 handled based on the following matrix:

$$97 \quad v_j^+ = \max\{v_{1j}, v_{2j}, \dots, v_{mj}\}, v_j^- = \min\{v_{1j}, v_{2j}, \dots, v_{mj}\} \tag{2}$$

99 Third, by the routine procedures we calculate the Euclidean distances S_i^+ and S_i^- between each selected
 100 indicator, as well as the positive-negative ideal solutions by the following Eq. (3):

$$102 \quad S_i^+ = \sqrt{\sum_{j=1}^n (v_{ij} - v_j^+)^2}, S_i^- = \sqrt{\sum_{j=1}^n (v_{ij} - v_j^-)^2} \tag{3}$$

103 Fourth, on this basis we further calculate the proximity of each sample city to the ideal solution, which herein is
 104 denoted as C_i representing the resilience index.:

$$105 \quad C_i = \frac{S_i^-}{S_i^+ + S_i^-} \tag{4}$$

106 where and a larger C_i indicates better assessment results of the object.

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108 **2.3 Data collection**

109 The data were obtained from the statistical yearbooks published by the selected sample city governments,
 110 China's population census, the water resources bulletins, the annual city reports posted on the website of
 111 meteorological department, press releases, and urban archive documents. By strictly selection and considering
 112 the data access, a sample of total 27 cities in China's seven major river basins was identified for this research.
 113 Although typhoon is categorized in magnitudes, such as tropical depression, tropical storm, severe typhoon and
 114 super typhoon, across the 27 sampled cities, there were no data for each category of typhoon after careful
 115 investigation. For this reason, we failed to classify the typhoon into subcategories by the magnitude. Apart from
 116 this, after process of standardizing for the data, the weights of assessment indicators were obtained by the
 117 Entropy-TOPSIS method, as here is omitted due to editor layout limits.

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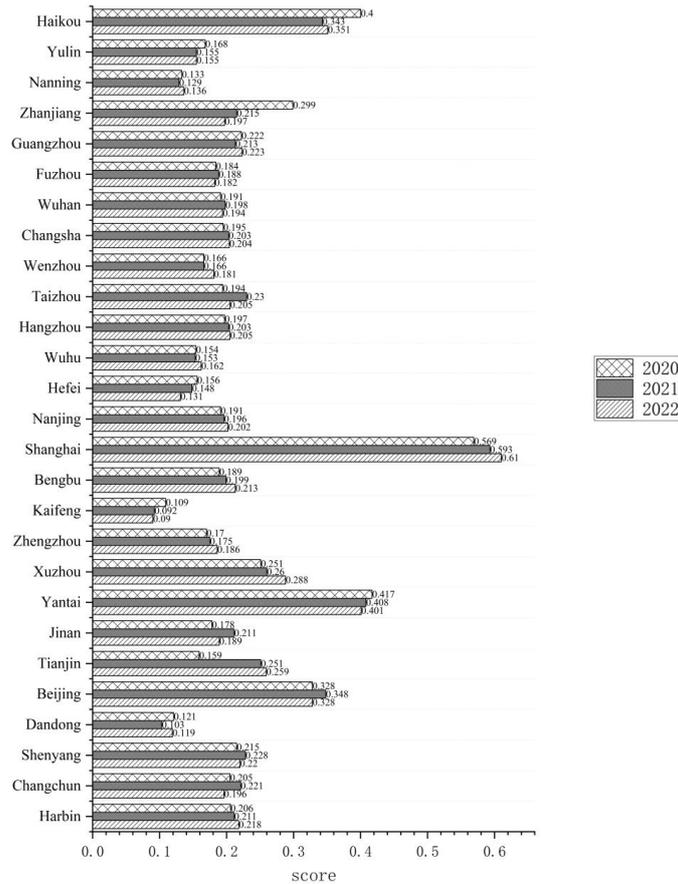
119 **3. Results and discussion**

120 **3.1 Spatial variation in urban resilience**

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122 Figure.1 shows the urban resilience level imposed by typhoons, and the sample cities within the seven river
123 basin could be categorized into four levels. The higher the index was, the weaker for the urban resilience to
124 external shocks of typhoons, which could also possibly lead to greater risks of derivative damages. In specific,
125 cities at *Level 1 to Level 3* are Shanghai (average 0.5907), Yantai (average 0.4087), Haikou and Beijing, respectively.
126 Cities at *Level 4* with a composite rating from 0.2 to 0.3, include Xuzhou, Zhanjiang, Tianjin, Shenyang,
127 Guangzhou, Harbin, Taizhou, Changchun, Hangzhou, Changsha, and Bengbu. *Level 5*, rating from 0.1 to 0.2,
128 include Nanjing, Wuhan, Jinan, Fuzhou, Zhengzhou, Wenzhou, Yulin, Wuhu, Hefei, Nanning, and Dandong.
129 *Level 6* is labeled to Kaifeng with a composite score of less than 0.1.

130 Generally, typhoons are affected by the geographical and meteorological factors. Therefore, it needs to examine
131 differences in similar settings. Thus, we classified the 27 key cities into seven regions by river basins where they
132 are geographically located, and further calculated resilience scores at each assessment dimension .



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Figure 1. Comprehensive assessment value of urban resilience of 27 selected cities in China.

Notes: The score denotes the urban resilience rating in a specific year. These scores are between 0 and 1. The higher the score is, the stronger resilient the city is to typhoon disaster.

137 **3.2. Inter-watershed comparison**

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The watershed scores were the arithmetic means of the assessment values in all typhoon-affected cities (Table 2). Specifically, urban resilience in the Haihe River Basin scores the highest value, as it encompasses the China's capital Beijing and the Metropolitan city of Tianjin. The Beijing-Tianjin region holds the advantages both in the



141 infrastructures and economic strength as well as a large number of high-level professional staffs in the fields of
 142 public health care, water conservancy, and a abundance of financial support from the central administration,
 143 which laid a solid foundation to enhancing the capacity of typhoon reduction. The resilience score of cities in the
 144 Nandu River Basin is the lowest by calculation. The reasons for this result can be summarized in multiple aspects.
 145 First, compared to others, the city of Haikou has a relatively weaker industrial level due to its more remote
 146 geographical distance from China’s major industrial bases or the economic delta, yielding a lower economic
 147 value of per capital GDP on average. Secondly, Haikou occupied a relatively high population density, which
 148 underscores the potential for risk exposure to typhoon disasters. Thirdly, the frequent cross-provincial outflow of
 149 population from Hainan to developed Guangdong province caused the loss of technical expertise, which was one
 150 of the obstacles for disaster reduction.

151 **Table 2** Dimensional ratings of 27 main typhoon-affected cities with respect to watersheds in China

| City | Score | Economic | Social | Environmental | Managerial |
|---|--------|----------|--------|---------------|------------|
| Songhua River Basin | | | | | |
| Harbin | 0.3336 | 0.2717 | 0.5143 | 0.3860 | 0.1623 |
| Changchun | 0.3446 | 0.3053 | 0.4390 | 0.2090 | 0.4250 |
| Basin Score | 0.3391 | | | | |
| Liaohe River Basin | | | | | |
| Shenyang | 0.4030 | 0.3343 | 0.8303 | 0.1733 | 0.2740 |
| Dandong | 0.3202 | 0.1753 | 0.8210 | 0.1990 | 0.0853 |
| Basin Score | 0.3616 | | | | |
| Haihe River Basin | | | | | |
| Beijing | 0.4755 | 0.7313 | 0.3923 | 0.2057 | 0.5727 |
| Tianjin | 0.4384 | 0.4303 | 0.3133 | 0.6220 | 0.3880 |
| Basin Score | 0.4570 | | | | |
| Yellow River -Huaihe River Basin | | | | | |
| Jinan | 0.3876 | 0.4730 | 0.5973 | 0.1643 | 0.3157 |
| Yantai | 0.3676 | 0.4140 | 0.5057 | 0.1710 | 0.3797 |
| Zhengzhou | 0.3185 | 0.3183 | 0.4357 | 0.1673 | 0.3527 |
| Xuzhou | 0.3192 | 0.3060 | 0.4237 | 0.0740 | 0.4730 |
| Kaifeng | 0.2157 | 0.2697 | 0.2677 | 0.0817 | 0.2437 |
| Bengbu | 0.3105 | 0.3160 | 0.4110 | 0.1613 | 0.3537 |
| Basin Score | 0.3198 | | | | |
| Yangtze River Basin | | | | | |
| Shanghai | 0.4462 | 0.7417 | 0.5230 | 0.1610 | 0.3590 |
| Nanjing | 0.3657 | 0.6253 | 0.4247 | 0.1713 | 0.2413 |
| Hefei | 0.3263 | 0.4003 | 0.4460 | 0.1867 | 0.2720 |
| Wuhu | 0.3482 | 0.4260 | 0.3933 | 0.1607 | 0.4127 |
| Hangzhou | 0.4164 | 0.5863 | 0.2303 | 0.4047 | 0.4443 |
| Taizhou | 0.2734 | 0.3953 | 0.3273 | 0.1810 | 0.1900 |
| Wenzhou | 0.2940 | 0.4590 | 0.1977 | 0.2043 | 0.3150 |
| Changsha | 0.3353 | 0.3780 | 0.4153 | 0.1233 | 0.4247 |
| Wuhan | 0.3798 | 0.5423 | 0.3383 | 0.1690 | 0.4697 |
| Fuzhou | 0.3313 | 0.4880 | 0.2860 | 0.1893 | 0.3620 |
| Basin Score | 0.3517 | | | | |
| Zhujiang River Basin | | | | | |
| Guangzhou | 0.2970 | 0.5763 | 0.1643 | 0.1667 | 0.2807 |
| Zhanjiang | 0.2428 | 0.3147 | 0.1743 | 0.1753 | 0.3067 |
| Nanning | 0.1877 | 0.2343 | 0.2473 | 0.1167 | 0.1523 |
| Yulin | 0.1970 | 0.2027 | 0.1863 | 0.0820 | 0.3170 |
| Basin Score | 0.2311 | | | | |
| Nandu River Basin | | | | | |
| Haikou | 0.2821 | 0.2917 | 0.3707 | 0.0310 | 0.4350 |
| Basin Score | 0.2821 | | | | |

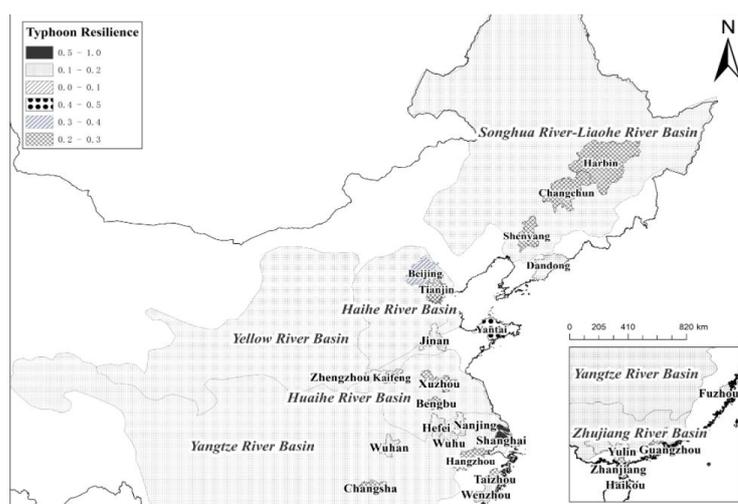
152 Comparatively, the resilience scores to typhoon in northeast region (Songhua & Liao River Basin) was slightly
 153 stronger than southeast region (Pearl River Basin). Although southeast coast region possesses advantages in
 154 economic strength and growth in comparison with the northeast, the former displayed a relatively lower
 155 proportion of investment in the field of reservoir construction, as constituted one of the principal indicators on



156 the economic dimension in our study. (assessment weight >10%). By governmental statistics, the water storage
 157 capacity of large and medium-sized reservoirs in the Northeast region of *Songhua* and *Liao* River Basins is
 158 significantly higher than that of other basins. Overall, the Pearl River Basin has a tropical and subtropical climate
 159 with an average annual temperature exceeding 22°C and precipitation exceeding 1,500 mm (Sheng, 2015), making
 160 them highly vulnerable to typhoons. Besides, as for the social dimension, the river basin of southeast with higher
 161 population density actually necessitated medical and rescue capacities in typhoon disaster, while the northeast
 162 basin is insufficient in the water facilities. For the environmental dimension, compared to the southeastern coastal
 163 regions with annual higher precipitation, the cities in northeast exhibited a lower risk of typhoons. Moreover, the
 164 calculated resilience value of the Yellow-Huai River Basin and Yangtze River Basin (including major cities in
 165 China's coaster provinces), is stronger than that of the Northeast. Both basins perform better in terms of social
 166 and administrative dimension, which can be attributed to the influx of R&D investment and government capacity
 167 (Ju, 2020). The Yellow-Huai River Basin is located in the eastern regions of China, with suitable temperatures and
 168 a high level of urbanization, which attracts a large inflow of talents and creates a better pool of specialized
 169 human resources in the fields of water conservancy and environment protection.

170 3.3. Intra-watershed comparison

171 China's administrative system endows a superior status either in economic and political aspects for provincial
 172 capitals, and these cities exhibit higher levels of urban resilience than other prefecture-level cities. Likewise these
 173 cities include Harbin, Nanjing, and Guangzhou, all of which have higher resilience scores than other types of
 174 cities in the province. Namely, we demonstrates that provincial capital cities are more resilient compared to the
 175 prefecture-level cities in confronting typhoon disasters. However, there are exceptions in some watersheds. For
 176 instance, Jinan, the capital city of Shandong Province, is slightly lower in resilience score than Yantai, a common
 177 prefecture-level city in Jiaodong Peninsula, which can be attributed to the following factors. On one side, Yantai
 178 was a city distinguished by economic prosperity thanks to industries and infrastructures. For economic strength
 179 and geographical location, Yantai administration puts emphasis on investments in typhoon-induced disaster
 180 prevention, with three to four times annual expenditures than that of Jinan, the capital city of Shandong Province.
 181 Secondly, Yantai was relatively low in the population density compared with Jinan, which was advantageous for
 182 the city when calculating per capital indicators.



183
184 **Figure 2.** Resilience rating of 27 typhoon-affected cities in respect to watersheds.

185 In the *Songhua* and *Liao* River Basin, cities perform poorly in economic and administrative dimensions but
 186 relatively better in social and environmental dimensions. Compared to Shenyang, the capital of Liaoning
 187 Province, Dandong has a weaker heavy industry base and a long coastline in Northeast China, but hold an
 188 advantage in terms of water conservancy conditions in geographical location. In the Songhua River basin, Harbin
 189 ranks 10th out of 27 cities in terms of typhoon resilience level, while Changchun ranks 12th. This difference may
 190 be closely related to the high population density in Changchun, and weaker reservoir storage capacity compared



191 to capital city of Harbin. For the whole northeast region in China, Harbin City has the largest reservoir capacity
192 (12,966 million cubic meters). Reservoirs play an important role in dissipating heavy precipitation caused by
193 typhoon in a short period of time and staggering the peaks. Besides, the Haihe River Basin presents the difference
194 in typhoon resilience between Beijing and other municipalities. In terms of typhoon resilience scores for 2020-
195 2022, Tianjin (0.2230) is significantly lower than Beijing (0.3347). Specifically, Beijing performs well in the
196 economic, social and administrative dimensions, with its water resources sector ranking first in terms of
197 information disclosure and disaster response.

198 In the Yellow-Huai River Basin, the observed city with the highest level of resilience is Jinan, which may be
199 closely related to its administrative location as the capital city of Shandong province and the command center of
200 the Liberation Army in China's Northern Theater of Operations. By the calculation result, Jinan scores high on the
201 social and administrative dimensions. Besides, Jinan has an advantage of personnel in typhoon disaster rescues,
202 including 27 water and environmental practitioners per 10,000 population in 2022, far more than any other city in
203 the basin. By comparison, as a provincial capital city in Henan province, Zhengzhou scored lower on the social
204 and administrative dimensions. The scale of inter-provincial population outflow in Henan Province is among the
205 highest in China, resulting in a lower level of technical level expertise for disaster reduction, which is one of the
206 barriers to enhancing resilience to typhoon disasters.

207 In the developed economic regions of Yangtze River Basin, Hangzhou City has the highest level of overall
208 resilience, mainly due to its high ranking in reservoir capacity and environmental management practitioners. The
209 leading development position in the Yangtze River Delta region gives the cities of Shanghai, Hangzhou, and
210 Nanjing a very high economic advantage, attracting high-level talents from all over the country to work and
211 settle, and provide abundant talents including disaster reduction of water conservancy and emergency aids.
212 Among the Yangtze River Basin, Taizhou City has the lowest level of resilience, which may be related to the city's
213 low reservoir capacity and lack of facilities to mitigate heavy rainfall caused by typhoons.

214 Cities in the Pearl River Basin are categorized into two levels, with two cities (Guangzhou, Zhanjiang) in
215 Guangdong Province rating the fourth status and one city (Nanjing) in Guangxi Province rating the fifth position.
216 The Pearl River Basin has abundant rainfall and is a typical rainy river with big river runoff, accounting for about
217 13% of the national runoff. The Xijiang and Beijiang rivers meet and have traditionally been the main source of
218 typhoon-induced flooding in the delta. Nanning and Yulin, although less pressurized in terms of population
219 density and large reservoirs of capacity, the level of water conservancy construction and disaster mitigation
220 demonstration communities is low, which leads to a weaker ability to confront typhoons. From an economic
221 perspective, the city of Nanning and Yulin was relatively laggard than that of other provincial capital cities,
222 particularly in terms of financial support. In terms of the environmental dimension, the capital city of Nanning
223 in Guangxi province encountered a highly frequent typhoon landfall, but the limited volume of local reservoirs
224 made it difficult to handle the erupt typhoon-induced rainfalls.

225 3.4. Temporal evolution of urban resilience to typhoon

226 By calculation, the metropolis of Shanghai ranked first in the comprehensive rating for confronting typhoon,
227 and the resilience score has been increasing every year (see Figure 1). Haikou, a city also with a high resilience
228 score, ranked steadily and possessed relatively high performance in social dimension. The decline of the overall
229 score in 2022 is related to the decrease in the number of disaster reduction model communities. The growth in the
230 density of drainage network in the built-up area of Haikou City demonstrates the government's determination to
231 improve the city's drainage capacity and to improve the mechanism for preventing typhoon disasters beforehand.
232 Besides, Xuzhou City, as the top-ranked city among the three tiers and a prefecture-level city in Jiangsu Province,
233 exceeds many provincial capitals in resilience scores, and its composite scores show an increasing tendency
234 annually. The spread of drainage network around the city and lifeline project enhanced its performance of
235 resilience to typhoon. Tianjin, the city with top overall score for resilience, improved its ranking from No. 22 in
236 2020 to No. 6 in 2022, with an average overall score of 0.2230, and the number of public information published
237 online by the Water Resources Bureau reflects the increased level of information capacity of typhoon disasters.

239 4. Conclusion

240 The brief communication reports an assessment framework for urban resilience to typhoon disaster and
241 applies in 27 cities in China's seven river basins based on watersheds during 2020–2022. Indicators with the
242 economic, social, environmental, and administrative dimensions were selected using the Entropy –TOPSIS
243 method. The resilience score to typhoons in these watersheds apparently differs and ranked as follows: *the Haihe*



244 *River Basin, the Liao River Basin, the Yangtze River Basin, the Songhua River Basin, the Yellow-Huai River Basin, the*
245 *Nandu River Basin and the Zhujiang River Basin.* We thus conclude in the following aspects:

246 (1) Level of economic power is not synchronized with resilience to typhoons. Overall, despite the economic
247 advantages in southeast, its resilience score is lower than that of the northeast. As some research indicate,
248 government expenditure and economic capacity in disaster reduction is positively correlated with the economic
249 losses when cities face typhoon disasters (Cutter, 2016; Hechanova, et al., 2018; Guo, 2020). Some cities with a
250 higher percentage of financial expenditure on typhoon and healthcare demonstrated a stronger capacity for
251 medical rescue and minimized casualties.

252 (2) In the environmental dimension, climatic condition, land resource, and the level of infrastructure are
253 important factors in affecting city's resilience capability to typhoon disaster. We found that some coaster cities
254 with higher precipitation annually scored high mainly thanks to its urban infrastructure and large-scale
255 facilities of water conservation, such as large and medium-sized reservoirs, hydropower station, waterway and
256 harbor projects. A typical city case was the Fuzhou in the coaster Fujian province. Although it received more
257 annual precipitation due to its location of subtropical monsoon, the reservoirs with large capacity and
258 infrastructure reduced risks of flood damages caused by typhoon.

259 (3) The administrative dimension with respect to resilience herein refers to the capacity of water conservancy
260 for local administration that better respond before, during, and after a city encounters typhoon. In this study, the
261 status of disaster reduction pilot communities was an important assessment indicator. With the help of
262 community construction, cities can strengthen the capacity of response on the grassroots level through effective
263 learning, and improve early warning measures for typhoon-induced disaster reduction. Besides, the government
264 ought to enhance the level of information disclosure of typhoon in the frontline water sector, for instance,
265 strengthening the information warning by adequate channels, such as government websites, weather forecasts,
266 and social platform.

267 To sum up, although we tried to make resilience assessment in a comprehensive dimensions to typhoons
268 based on complexity of China's geographical river basin context, there are still more gaps awaiting filled. More
269 cities of different types and the factor of heterogeneity are supposed to be added with a dynamic perspective.
270 Also, we could conduct future's research aiming at conceptualizing cascading impacts. of typhoon disaster to
271 further explore resilience assessment.

272

273 **Data availability**

274 The data set is available on request.

275 **Author contribution**

276 **Liu:** conceptualization, analysis, writing – original draft, and funding acquisition.

277 **Yang:** data curation, writing – original draft, **Wu:** visualization, writing – review, and editing

278 **Declaration of competing interest**

279 The authors declare that neither of the authors has any competing interests.

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