

# Removal Effect of Precipitation on Atmospheric Particulate Matter (PM<sub>2.5</sub>) in Bengbu City

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**Abstract** Based on the monitoring data of PM<sub>2.5</sub> concentration in Bengbu Environmental Monitoring Station and precipitation observation data of Bengbu National Meteorological Observation Station from 2016 to 2019, the influence of precipitation on PM<sub>2.5</sub> mass concentration in Bengbu City was analyzed. The results show that precipitation had a washing and removal effect on PM<sub>2.5</sub> in the air, and the removal effect was related to precipitation level, precipitation intensity, precipitation duration and PM<sub>2.5</sub> concentration. The removal effect of precipitation on PM<sub>2.5</sub> increased with the increase of precipitation level, and the seasonal difference was obvious. Precipitation intensity was positively correlated with the removal effect of PM<sub>2.5</sub>, but the average removal rate began to decline when precipitation intensity exceeded 10 mm. With the increase of precipitation intensity, the proportion of positive removal showed an overall upward trend, but there was a low-value area as precipitation intensity was 3 – 10 mm. Precipitation duration was also positively correlated with the removal effect of PM<sub>2.5</sub>, and there was a low-value area when precipitation duration was 10 – 15 h. When PM<sub>2.5</sub> concentration was low before the precipitation process began, the removal effect was not good, and the average removal rate was negative. As PM<sub>2.5</sub> concentration was high before the precipitation process started, the removal effect was obvious.

**Key words** Precipitation; PM<sub>2.5</sub>; Removal efficiency; Bengbu

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PM<sub>2.5</sub>, also known as fine particulate matter, refers to aerosol particles with a diameter less than or equal to 2.5 μm in the atmosphere, and can be suspended in the air for a long time. Due to its small particle size, large area and strong activity, it is easy to carry toxic and harmful substances, and has a great impact on atmospheric environmental quality and human health<sup>[1]</sup>. Since the 1990s, with the acceleration of urbanization, atmospheric environmental pollution events have increased significantly, and their impact on economic and social development has been increasingly intensified<sup>[2]</sup>.

China is the world's largest population country and a developing country. China's economy is in a period of rapid development. With the increasing demand for energy, the rapid expansion of urban population, and the emergence of a large number of industrial and mining enterprises, all kinds of pollution into the atmosphere have made air pollution more and more serious. Many domestic research results show that the proportion of fine particulate matter in China's air is on the rise, and fine particulate matter has become the main air pollutant affecting all regions. How to improve the quality of urban air environment and how to balance environment and development have become major issues that need to be solved urgently.

Based on the daily monitoring data of air pollution, surface meteorological observation data and radiosonde data in 120 key cities for environmental protection in China, Zhang Ying *et al.*<sup>[3]</sup> analyzed the air quality of typical cities in China and its correlation with pollution meteorological parameters. The results show that air

quality index (AQI) in typical Chinese cities decreased from north to south; it was high in winter and low in summer in northern cities, and the variation was significantly greater than that in southern cities. Li Wenjie *et al.*<sup>[4]</sup> studied the spatial and temporal distribution characteristics of air pollution index (API) and its relationship with meteorological elements in Beijing, Tianjin and Shijiazhuang. Jiang Di *et al.*<sup>[5]</sup> found that there was a close relationship between PM<sub>2.5</sub> concentration and meteorological elements such as precipitation, wind speed, mixed layer thickness and relative humidity. In addition, the temporal and spatial distribution characteristics of particulate pollutants (PM<sub>2.5</sub> and PM<sub>10</sub>) in Beijing, Shanghai, Shijiazhuang, Xi'an and Wuhan and their relationship with meteorological conditions were analyzed<sup>[6–10]</sup>. The removal effect of precipitation on particulate matter in Beijing, Wuxi, Guiyang and Hefei was studied<sup>[11–14]</sup>.

At present, the domestic research on fine particulate pollutants (PM<sub>2.5</sub>) is mostly concentrated in the Beijing – Tianjin – Hebei, Yangtze River Delta and other economically developed regions or a few central cities, and the research on small and medium-sized cities in Anhui Province is relatively rare. In this paper, the relationship between precipitation and PM<sub>2.5</sub> concentration in Bengbu was studied, and the removal effect of precipitation of different levels and intensities on PM<sub>2.5</sub> was explored, so as to provide reference for the monitoring, early warning and pollution prevention of PM<sub>2.5</sub> in cities.

## 1 Data and methods

**1.1 Data sources** The meteorological data used in this paper were the daily and hourly observation data of precipitation in

Bengbu National Meteorological Observation Station (No. 58221) from January 1, 2016 to December 31, 2019, with a temporal resolution of 1 d and 1 h, respectively.

The daily and hourly data of PM<sub>2.5</sub> concentration from January 1, 2016 to December 31, 2019 came from six urban environmental monitoring stations in Bengbu, including Department Store, the Second Water Plant (adjusted to Tianjiabing Middle School in 2018), Bengbu College, High-tech Zone, Workers' Sanatorium and government of Huaishang District, with a temporal resolution of 1 d and 1 h, respectively. Missing test data were excluded before statistical analysis.

**1.2 Methods** Daily removal efficiency ( $RF$ ) is calculated by the particle concentration on the day before the precipitation starts ( $CON1$ ) and the particle concentration on the precipitation day ( $CON2$ ) as follows:

$$RF = (CON1 - CON2) / CON1 \times 100\%$$

$RF > 0$  means positive removal, while  $RF < 0$  means negative removal, and  $RF = 0$  is zero removal.

The removal process was extracted and analyzed according to the rainfall process. The first occurrence of precipitation above 0.1 mm was as the start of the process, and the last occurrence of precipitation above 0.1 mm was as the end of the process.

The 1-h average particle concentration ( $CON1$ ) before the start of the process was defined as the particle concentration before the removal process, and the 1-h average particle concentration ( $CON2$ ) after the end of the process was defined as the particle concentration after the removal process. The average removal efficiency of  $n$  processes was calculated by the average concentration of aerosol particles before and after  $n$  processes:

$$\overline{RF} = \frac{\sum_{i=1}^n (CON1 - CON2)}{\sum_{i=1}^n CON1} \times 100\%$$

The removal process with  $RF > 0$  is a positive removal process, while the removal process with  $RF < 0$  is a negative removal process, and the removal process with  $RF = 0$  is a zero removal process. The percentage of positive removal processes is the ratio of the number of positive removal processes to the total number of removal processes.

Air quality is divided into six levels according to the *Technical Regulation on Ambient Air Quality Index* (on trial) (HJ 633 – 2012) (Table 1).

**Table 1** Air quality class and concentration of atmospheric particulate matter

Air quality	AQI	μg/m <sup>3</sup>	
		PM <sub>2.5</sub> mass concentration	PM <sub>10</sub> mass concentration
Excellent	0 – 50	0 – 35	0 – 50
Good	51 – 100	35 – 75	50 – 150
Light pollution	101 – 150	75 – 115	150 – 250
Moderate pollution	151 – 200	115 – 150	250 – 350
Heavy pollution	201 – 300	150 – 250	350 – 420
Serious pollution	> 300	> 250	> 420

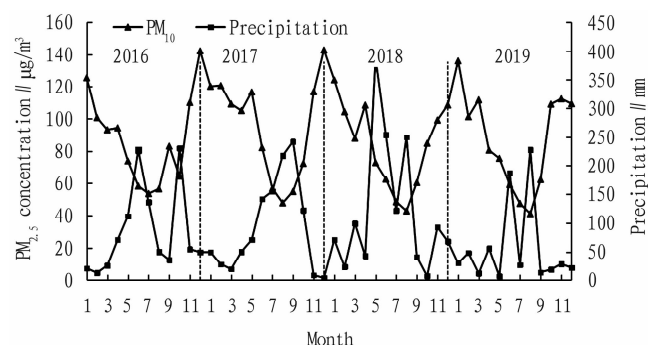
On the basis of the definition of 24-h precipitation in the

*Classification Standard for Precipitation Intensity (Inland Part)* of China Meteorological Bureau, rainfall was divided into four grades: light rain (0.1 – 9.9 mm), moderate rain (10 – 24.9 mm), heavy rain (25 – 49.9 mm), as well as rainstorm and above ( $\geq 50$  mm).

## 2 Distribution characteristics of precipitation and PM<sub>2.5</sub> concentration

Fig. 1 shows the monthly precipitation and average PM<sub>2.5</sub> concentration of Bengbu City from 2016 to 2019. The annual average precipitation of Bengbu City was 1 066.7 mm, and the annual average precipitation days were 105.8 d. According to the 30-year climate data from 1980 to 2010 released by the National Climate Center, the annual average precipitation in Bengbu City was 965.9 mm, and the annual average precipitation days were 102.4 d. Therefore, the precipitation in Bengbu City from 2016 to 2019 was normal.

Precipitation has a washing and removal effect on pollutants in the air, especially particulate matter. The precipitation of Bengbu City from 2016 to 2019 was more from May to August, but less from December to February. Correspondingly, PM<sub>2.5</sub> concentration was the lowest during July – August and highest in December. Precipitation had a significant negative correlation with PM<sub>2.5</sub> concentration, passing 99% confidence interval test. The monthly mean mass concentration of particulate matter was lower in months with more precipitation days and precipitation. In months with fewer precipitation days and less precipitation, the monthly mean mass concentration of particulate matter was higher.



**Fig. 1** Monthly precipitation and PM<sub>2.5</sub> concentration in Bengbu City from 2016 to 2019

## 3 Removal effect of precipitation on PM<sub>2.5</sub>

**3.1 Influence of precipitation grade on the removal** Fig. 2 shows the PM<sub>2.5</sub> concentration under different levels of precipitation in Bengbu City from 2016 to 2019. It can be seen from the figure that generally, with the increase of precipitation level, PM<sub>2.5</sub> concentration declined, but the difference between seasons was obvious. The removal effect was most obvious in winter, followed by spring and autumn, and less obvious in summer. In winter, the average concentration of PM<sub>2.5</sub> on light, moderate and heavy rain days were 87.70, 41.33 and 45.50 μg/m<sup>3</sup>, respectively. The removal rate of light rain, moderate rain and heavy rain were

−14% , 31% and 37% , respectively. It can be seen that the washing effect of precipitation on  $PM_{2.5}$  became more and more significant with the increase of precipitation level.

$PM_{2.5}$  concentration and removal rate on different grades of rain days in autumn and spring were lower than that in winter. The removal effect of light rain and moderate rain on  $PM_{2.5}$  in summer significantly reduced. The average concentration of  $PM_{2.5}$  on light, moderate and heavy rain days in summer was 42.38, 23.40 and 38.00  $\mu g/m^3$ , respectively, and the removal rate was −24% , 15% and 6% , respectively. In summer,  $PM_{2.5}$  concentration even increased, and the removal rate decreased on heavy rain days. In Bengbu City,  $PM_{2.5}$  concentration was the lowest from June to August. The phenomenon that  $PM_{2.5}$  concentration rose and the removal rate declined on heavy rain days may be related to the low initial concentration of  $PM_{2.5}$ .

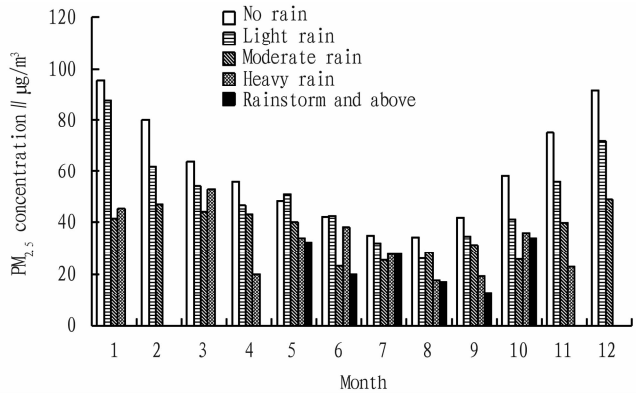


Fig.2  $PM_{2.5}$  concentration on different levels of precipitation days in Bengbu City from 2016 to 2019

3.2 Influence of precipitation intensity on the removal effect

According to the classification of hourly precipitation in Bengbu City from 2016 to 2019 according to precipitation intensity, it is found that the removal effect of  $PM_{2.5}$  had a good relation with precipitation intensity, and the removal effect became more obvious with the increase of precipitation intensity. As can be seen from Fig. 3, with the increase of precipitation intensity, the average removal rate became positive, and the upward trend was obvious. When precipitation intensity exceeded 10 mm, the average removal rate began to decline. With the increase of precipitation intensity, the proportion of positive removal showed an overall upward trend, but decreased slightly when precipitation intensity was 3 – 10 mm.

As can be seen from Table 2, when precipitation intensity was less than 1 mm, the removal effect of precipitation on  $PM_{2.5}$  was not obvious. The proportion of positive and negative removal was equal, and the average removal rate was negative. When precipitation intensity increased to 1 – 3 mm, the removal effect of precipitation on  $PM_{2.5}$  was significantly improved, and the proportion of positive and negative removal increased to 68.22%. The average removal rate changed from negative to positive value, increasing to 3.25%. As precipitation intensity increased to 3 – 10 mm, the removal effect of precipitation on  $PM_{2.5}$  was improved, but the change was not obvious, and the proportion of positive re-

moval decreased slightly to 65.29% , while the average removal rate increased significantly to 9.20% . When precipitation intensity increased to more than 10 mm, the removal effect of precipitation on  $PM_{2.5}$  was not significantly improved, and the proportion of positive removal further increased to 69.64% , but the average removal rate reduced to 5.49% .

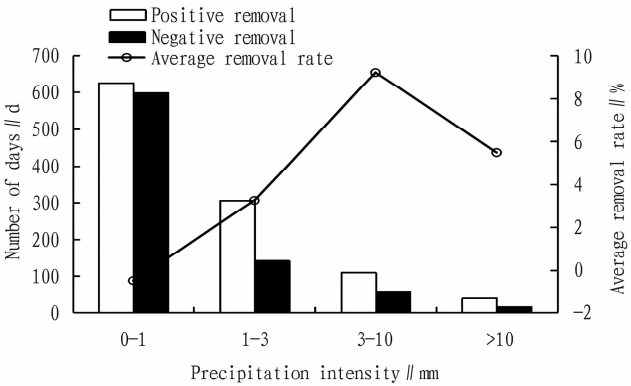


Fig.3 Removal of  $PM_{2.5}$  by precipitation of different intensity in Bengbu City from 2016 to 2019

Rainfall of more than 5 mm in Bengbu City mostly occurred from May to October, while the  $PM_{2.5}$  concentration in Bengbu City from May to October was relatively low in the whole year. When precipitation intensity was 3 – 10 mm, the proportion of positive removal decreased slightly; as precipitation intensity exceeded 10 mm, the average removal rate decreased, which may be affected by the low  $PM_{2.5}$  concentration before the precipitation started, which will be analyzed in detail below.

Table 2 Relationship between precipitation intensity and the removal rate of  $PM_{2.5}$  in Bengbu City from 2016 to 2019

Precipitation intensity // mm	Number of positive removal samples	Number of negative removal samples	Average removal rate // %
0 – 1	625	598	−0.50
1 – 3	307	143	3.25
3 – 10	111	59	9.20
> 10	39	17	5.49

3.3 Effects of the duration of precipitation on the removal effect

By classifying the precipitation in Bengbu City from 2016 to 2019 according to the duration of precipitation, it is found that there was a certain relationship between the removal effect of  $PM_{2.5}$  and the duration of precipitation. As can be seen from Table 3, when the duration of precipitation was 1 h, the removal effect of precipitation on  $PM_{2.5}$  was not obvious. The number of positive and negative removals was the same, and the average removal rate was negative. As the duration of precipitation was 2 – 3 h, the removal effect of precipitation on  $PM_{2.5}$  was improved, and the proportion of positive removal was enhanced. The average removal rate turned positive. When the duration of precipitation increased to 4 – 9 h, the removal effect of precipitation on  $PM_{2.5}$  steadily increased, and the proportion of positive removal reached 60% , while the average removal rate decreased slightly. As the duration of precipitation increased to 10 – 15 h, the removal effect of pre-

cipitation on PM<sub>2.5</sub> declined, and both positive removal and average removal rate decreased. The longer duration of precipitation mostly occurred from May to September, while the PM<sub>2.5</sub> concentration from May to September was low in the whole year. This decrease in the removal rate may be related to the PM<sub>2.5</sub> concentration before precipitation. When the duration of precipitation exceeded 16 h, the removal effect of precipitation on PM<sub>2.5</sub> was greatly improved, and the negative removal only existed sporadically. It can be seen that in general, the longer the duration of precipitation, the better the removal effect, the higher the positive removal rate.

**Table 3** Relationship between the duration of precipitation and the removal rate of PM<sub>2.5</sub> in Bengbu City from 2016 to 2019

Duration of precipitation//h	Number of positive removal samples	Number of negative removal samples	Proportion of positive removal//%
1	235	227	50.87
2–3	270	212	56.02
4–9	379	242	61.03
10–15	123	92	57.21
>16	88	17	83.81

**3.4 Effect of initial PM<sub>2.5</sub> concentration on the removal effect** By analyzing the relationship between 648 precipitation processes and PM<sub>2.5</sub> concentration in Bengbu City from January 1, 2016 to December 31, 2019, it is found that the removal effect of precipitation on PM<sub>2.5</sub> was not only related to the time and intensity of precipitation, but also had a great relationship with PM<sub>2.5</sub> concentration before precipitation.

Seen from Table 4, when the PM<sub>2.5</sub> concentration before the precipitation process was less than 75 μg/m<sup>3</sup> (air quality was excellent or good), the removal effect of precipitation on PM<sub>2.5</sub> was negative, and the average removal rate was −3.56%, with little difference between positive and negative removal. As the PM<sub>2.5</sub> concentration before the precipitation process was 76–150 μg/m<sup>3</sup> (light and moderate pollution), the average removal rate turned positive, rising significantly to 13.99%, and the proportion of positive removal rose significantly. When the PM<sub>2.5</sub> concentration before the precipitation process was greater than 150 μg/m<sup>3</sup> (heavy pollution and above), the average removal rate rose to 34.58%, and all removal was positive. It can be seen that the removal effect of precipitation on PM<sub>2.5</sub> was closely related to the PM<sub>2.5</sub> concentration before the precipitation process. When PM<sub>2.5</sub> concentration was low, the removal effect of precipitation on PM<sub>2.5</sub> was not good, and the average removal rate was negative. However, as PM<sub>2.5</sub> concentration was high, precipitation had an obvious removal effect on PM<sub>2.5</sub>.

**Table 4** Relationship between the initial concentration and removal rate of PM<sub>2.5</sub> in Bengbu City from 2016 to 2019

Initial concentration//μg/m <sup>3</sup>	Removal rate during the process//%	Number of positive removal samples	Number of negative removal samples
0–75	−3.56	281	268
76–150	13.99	55	35
>150	34.58	9	0

4 Conclusions

- (1) Overall, the removal effect of precipitation on PM<sub>2.5</sub> rose with the increase of precipitation level, and the seasonal difference was obvious. The removal effect was most obvious in winter, followed by spring and autumn, while it was less obvious in summer.
- (2) The removal effect of precipitation on PM<sub>2.5</sub> was greatly affected by precipitation intensity. As a whole, the removal effect was enhanced with the increase of precipitation intensity. However, when precipitation intensity exceeded 10 mm, and the average removal rate began to decline. With the increase of precipitation intensity, the proportion of positive removal also showed an upward trend, but there was a low-value area when precipitation intensity was 3–10 mm.
- (3) Another important factor influencing the removal effect of precipitation on PM<sub>2.5</sub> was the duration of precipitation. In general, the longer the duration of precipitation, the better the removal effect, and the higher the positive removal rate. However, when the duration of precipitation was 10–15 h, the removal effect had a low-value area.
- (4) The removal effect of precipitation on PM<sub>2.5</sub> was also affected by the concentration of PM<sub>2.5</sub> before precipitation. When PM<sub>2.5</sub> concentration was low before the process began, the removal effect was not good, and the average removal rate was negative. However, as PM<sub>2.5</sub> concentration was high before the process started, the removal effect was obvious.

References

[1] WANG HC, WU ZB, ZHOU JB, *et al.* Relationship between PM<sub>2.5</sub> concentration and meteorological elements at Shangdianzi station of Beijing [J]. *Journal of Meteorology and Environment*, 2015, 31(5): 99–104.

[2] LI Y. The space-time variations of PM<sub>10</sub> concentration in major cities of China during 2000–2007[J]. *Journal of Arid Land Resources and Environment*, 2009, 23(9): 51–54.

[3] ZHANG Y, JIA XW, YANG X, *et al.* Characteristics of air pollution and its relationship with meteorological parameters in typical representative cities of China[J]. *Journal of Meteorology and Environment*, 2017, 33(2): 70–79.

[4] LI WJ, ZHANG SH, GAO QX, *et al.* Relationship between temporal-spatial distribution pattern of air pollution index and meteorological elements in Beijing, Tianjin and Shijiazhuang [J]. *Resources Science*, 2012, 34(8): 1392–1400.

[5] JIANG D, LI C. Relationship of the diffusion of PM<sub>2.5</sub> and meteorological conditions in Nanjing urban area[J]. *The Administration and Technique of Environmental Monitoring*, 2016, 28(1): 36–40.

[6] ZHOU YM, ZHAO XY. Correlation analysis between PM<sub>2.5</sub> concentration and meteorological factors in Beijing area[J]. *Acta Scientiarum Naturalium Universitatis Pekinensis*, 2017, 53(1): 111–124.

[7] GU KH, SHI HX, ZHANG S, *et al.* Variation characteristics of PM<sub>2.5</sub> levels and the influence of meteorological conditions on Chongming Island in Shanghai [J]. *Resources and Environment in the Yangtze Basin*, 2015, 24(12): 2108–2116.

[8] HAN JC, CHEN J, QIAN WM, *et al.* The research on relationship between meteorological condition and atmospheric particles in Shijiazhuang [J]. *Environmental Monitoring in China*, 2016, 32(2): 31–37.

[9] LI HX, SHI XM. Temporal and spatial distribution, meteorological factors of PM<sub>2.5</sub> in Xi'an City [J]. *Ecology and Environmental Sciences*, 2016, 25(2): 266–271.

- vironmental Science, 2022, 42(4): 1963–1974.
- [10] PENG J, DANG WX, LIU YX, *et al.* Review on landscape ecological risk assessment[J]. *Acta Geographica Sinica*, 2015, 70(4): 664–677.
  - [11] YANG F, JIN XB, LIU J, *et al.* Assessing landscape ecological risk in rapidly urbanized areas from the perspective of spatiotemporal dynamics[J]. *Transactions of the Chinese Society of Agricultural Engineering*, 2023, 39(18): 253–261.
  - [12] XU W, WANG J, ZHANG M, *et al.* Construction of landscape ecological network based on landscape ecological risk assessment in a large-scale opencast coal mine area[J]. *Journal of Cleaner Production*, 2021, 286: 125523.
  - [13] ZHANG Z, GONG J, PLAZA A, *et al.* Long-term assessment of ecological risk dynamics in Wuhan, China: Multi-perspective spatiotemporal variation analysis[J]. *Environmental Impact Assessment Review*, 2024, 105: 107372.
  - [14] WANG JL, CHEN CL, NI JP, *et al.* Resistance evaluation and "source-sink" risk spatial pattern of agricultural non-point source pollution in small catchment[J]. *Transactions of the Chinese Society of Agricultural Engineering*, 2018, 34(10): 216–224, 306.
  - [15] WANG X, SUN Y, LIU Q, *et al.* Construction and optimization of ecological network based on landscape ecological risk assessment: A case study in Jinan[J]. *Land*, 2023, 12: 743.
  - [16] LEUVEN RSEW, POUDDEVIGNE I. Riverine landscape dynamics and ecological risk assessment: Riverine landscape dynamics[J]. *Freshwater Biology*, 2002, 47: 845–865.
  - [17] DU L, DONG C, KANG X, *et al.* Spatiotemporal evolution of land cover changes and landscape ecological risk assessment in the Yellow River basin, 2015–2020[J]. *J. Environ. Manage.*, 2023, 332: 117149.
  - [18] LIANG T, DU P, YANG F, *et al.* Potential land-use conflicts in the urban center of Chongqing based on the "production-living-ecological space" perspective[J]. *Land*, 2022, 11: 1415.
  - [19] WU J, ZHU, Q, QIAO N, *et al.* Ecological risk assessment of coal mine area based on "source-sink" landscape theory: A case study of Ping-shuo mining area[J]. *J. Clean Prod.*, 2021, 295: 126371.
  - [20] CHEN XQ, DING ZY, YANG J, *et al.* Ecological risk assessment and driving force analysis of landscape in the compound mine-urban area of the northern Peixian County[J]. *Chinese Journal of Ecology*, 2022, 41(9): 1796–1803.
  - [21] LI J, PU R, GONG H, *et al.* Evolution characteristics of landscape ecological risk patterns in coastal zones in Zhejiang Province, China[J]. *Sustainability*, 2017, 9: 584.
  - [22] YUAN BY, GAO JH, CHI Y, *et al.* Cross-scale spatiotemporal characteristics of landscape ecological conditions index in coastal zone of Jiangsu Province, China during 1990–2020[J]. *Chinese Journal of Applied Ecology*, 2022, 33(2): 489–499.
  - [23] ZHANG Y, ZHANG F, WANG J, *et al.* Analysis of the temporal and spatial dynamics of landscape patterns and hemeroby index of the Ebinur Lake Wetland Nature Reserve, Xinjiang over the last 40 years[J]. *Acta Ecologica Sinica*, 2017, 37(21): 7082–7097.
  - [24] WANG H, LIU X, ZHAO C, *et al.* Spatial-temporal pattern analysis of landscape ecological risk assessment based on land use/land cover change in Baishuijiang National Nature Reserve in Gansu Province, China[J]. *Ecological Indicators*, 2021, 124: 107454.
  - [25] ZHAO FF, SUN GL, JI XM, *et al.* Land landscape pattern and ecological risk analysis of Tianshan Grand Canyon National Forest Park[J]. *Xinjiang Agricultural Sciences*, 2022, 59(3): 735–743.
  - [26] LIU J, WANG M, YANG L. Assessing landscape ecological risk induced by land-use/cover change in a county in China: A GIS-and landscape-metric-based approach[J]. *Sustainability*, 2020, 12: 9037.
  - [27] HUANG J, HU Y, ZHENG F. Research on recognition and protection of ecological security patterns based on circuit theory: A case study of Jinan City[J]. *Environ. Sci. Pollut. Res.*, 2020, 27: 12414–12427.
  - [28] WU L, ZHOU ZF, ZHANG L, *et al.* Response of landscape pattern change to habitat quality in karst area[J]. *Environmental Science & Technology*, 2023, 46(9): 206–217.
  - [29] LI C, ZHANG J, PHILBIN SP, *et al.* Evaluating the impact of highway construction projects on landscape ecological risks in high altitude plateaus[J]. *Sci Rep*, 2022, 12: 5170.
  - [30] JI Y, BAI Z, HUI J. Landscape ecological risk assessment based on LUCC: A case study of Chaoyang County, China[J]. *Forests*, 2021, 12: 1157.
  - [31] ZHANG XB, SHI PJ, LUO J, *et al.* The ecological risk assessment of arid inland river basin at the landscape scale: A case study on Shiyang River basin[J]. *Journal of Natural Resources*, 2014, 29(3): 410–419.
  - [32] JIN X, JIN Y, MAO X. Ecological risk assessment of cities on the Tibetan Plateau based on land use/land cover changes: Case study of Delingha City[J]. *Ecological Indicators*, 2019, 101: 185–191.
  - [33] ZHANG H, XUE L, WEI G, *et al.* Assessing vegetation dynamics and landscape ecological risk on the mainstream of Tarim River, China[J]. *Water*, 2020, 12: 2156.
  - [34] TAO C, ANMING B, HAO G, *et al.* Ecological vulnerability assessment for a transboundary basin in Central Asia and its spatiotemporal characteristics analysis: Taking Amu Darya River basin as an example[J]. *Journal of Natural Resources*, 2019, 34: 2643.
  - [35] ZHANG JC, GAO P, DONG XD, *et al.* Ecological vulnerability assessment of Qingdao coastal zone based on landscape pattern analysis[J]. *Journal of Ecology and Rural Environment*, 2021, 37(8): 1022–1030.
  - [36] XI SJ, CAI PL, AN YL. Spatio-temporal change and driving factors of integrated ecological risk of catchments in karst mountainous area of Guizhou Province from 2000 to 2018[J]. *Journal of Ecology and Rural Environment*, 2020, 36(9): 1106–1114.
  - [37] TONG X, WANG K, BRANDT M, *et al.* Assessing future vegetation trends and restoration prospects in the karst regions of southwest China[J]. *Remote Sens.*, 2016, 8: 357.
  - [38] WANG X, WANG R. Terrain gradient effect of land use and its driving factors in the Qinghai-Tibet Plateau[J]. *Pol. J. Environ. Stud.*, 2022, 31: 5299–5312.
  - [39] WANG JY, GUAN YH, WU XQ. Evolution of landscape ecological risk and its topographic differentiation in Karst Fault Basin[J]. *Acta Ecologica Sinica*, 2023, 43(19): 8167–8180.
  - [40] ZHANG XD, ZHAO ZP, ZHAO YX, *et al.* Landscape ecological risk assessment and ecological security pattern optimization construction in Yinchuan City[J]. *Arid Land Geography*, 2022, 45(5): 1626–1636.
  - [41] CHENG Y, GUAN YH, WU XQ. The spatial and temporal evolution of landscape pattern and landscape ecological security assessment in karst fault basin based on land use change[J]. *Acta Ecologica Sinica*, 2023, 43(22): 9471–9485.

(From page 15)

- [10] HUANG YL, LIU C, ZENG KF, *et al.* Spatio-temporal distribution of PM<sub>2.5</sub> in Wuhan and its relationship with meteorological condition, 2013–2014[J]. *Ecology and Environmental Sciences*, 2015, 24(8): 1330–1335.
- [11] WU J, SUN ZB, ZHAI L, *et al.* Effects of different precipitation types on aerosol particles in Beijing[J]. *China Environmental Science*, 2018, 38(3): 812–821.
- [12] ZHOU B, LIU DY, WEI JS, *et al.* A preliminary analysis on scavenging effect of precipitation on aerosol particles[J]. *Resources and Environment in the Yangtze Basin*, 2015, 24(Z1): 160–170.
- [13] DUAN Y, WU ZP, ZHANG DH, *et al.* Wet scavenging effect of precipitation on PM<sub>2.5</sub> pollutant in Guiyang[J]. *Meteorological Science and Technology*, 2016, 44(3): 458–462.
- [14] YU CX, DENG XL, SHI CE, *et al.* The scavenging effect of precipitation and wind on PM<sub>2.5</sub> and PM<sub>10</sub>[J]. *Acta Scientiae Circumstantiae*, 2018, 38(12): 4620–4629.