

Article

The Reducing Effect of Green Spaces with Different Vegetation Structure on Atmospheric Particulate Matter Concentration in Baoji City, China

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Abstract: With the acceleration of urbanisation and industrialisation, atmospheric particulate pollution has become one of the most serious environmental problems in China. In this study, green spaces in Baoji city were classified into different patterns on the basis of vegetation structural parameters, i.e., horizontal structure, vertical structure and vegetation type. Eleven types of green space with different structures were selected for investigating the relationships between atmospheric particulate matter (PM) concentration and green spaces with different vegetation structure, based on the “matrix effect” of environmental factors, i.e., location, time, wind velocity, temperature, humidity and area to the concentration of PM_{2.5} and PM₁₀ in the green spaces. The results showed that: (1) Location, time, wind velocity, temperature and humidity had highly significant effects on the concentration of PM_{2.5} and PM₁₀. In sunny and breeze weather conditions, PM_{2.5} and PM₁₀ concentration increased with the wind velocity and humidity, and decreased with the temperature. The range of PM₁₀ concentration was greater than the range of PM_{2.5} concentration. (2) Less than 2 hectares of the green space had no significant influence on the concentration of PM_{2.5} and PM₁₀. (3) The concentration of PM_{2.5} and PM₁₀ showed no significant difference between all the green spaces and the control group. There was no significant difference in the reduction of PM_{2.5} concentration between different structural green spaces, but there was a significant difference in the reduction of PM₁₀ concentration. The above results will provide a theoretical basis and practical methods for the optimisation of urban green space structures for improving urban air quality effectively in the future.

Keywords: PM_{2.5}; PM₁₀; city; meteorological factors; green space

1. Introduction

With the rapid development of urbanisation and industry in China and the increase of the per capita holdings of vehicles, the pollution of atmospheric particulate matter has become one of the most serious problems that cannot be neglected in today’s society [1,2]. Particulate matter is usually divided into four categories according to the aerodynamic diameter (D_p), and the effects of particulate matter on health vary with the different size fractions. PM₁₀ is one of the primary pollutants that affect air quality and damages human health through the respiratory tract [3]. Smaller PM_{2.5} is even more harmful, more easily enriches toxic substances, and can enter the alveolar and the blood circulation system, causing a variety of human systemic diseases [4–6]. In the 74 leading cities of China, approximately 32% of the reported deaths, with a mortality rate of 1.9%, were associated with PM_{2.5} and PM₁₀ in 2013, in which deaths from cardiovascular, respiratory and lung-cancer causes accounted

for 20% of the reported deaths, with a mortality rate of 1.2% [3]. Other studies have claimed that PM_{2.5} has contributed as much as 40.3% to total stroke deaths in China since 2015 [7]. Air pollution control is thus urgently needed in China.

Urban green spaces, as an important part of the city, not only provide people with recreational places [8,9], but also play a great role in regulating the urban ecological environment [10,11], especially in purifying and improving the air quality [12,13]. Many studies have claimed that plants could absorb atmospheric particles due to their special leaf surface structures and physiological and biochemical characteristics [14–16]. Vegetation structure is often used to describe the spatial distribution of vegetation in green space. The capacity of different vegetation structures to reduce the concentration of airborne particulates is different. Some studies have shown that the reduction of airborne particulate matter of composite structure of green space was better than the single structure of green space [17,18]. The concentration of particulate matter was lowest in the broadleaf and coniferous mixed-trees type and highest in the grassland type [19]. In addition, the reduction of PM₁₀ of mixed forest was better than the single forest [20]. However, other studies have shown that the ability to reduce the concentration of PM_{2.5} in the air of lawn and pure coniferous forest was stronger than that of the composite structure [21]. In the study of Yin [22], it was found that when the classification of vegetation structure was based on the overall spatial structure, the percentage of green space purification to TSP was positively correlated with the canopy density of the plant community, and it was negatively correlated with the permeability. Different plant allocation of a green space showed no significant difference in terms of the reducing effect on air particulate matter concentration [23]. It can be explained that there is no uniform vegetation structural classification standard and that ignoring the “background effect” of environmental meteorological factors are the main factors restricting the relationship between green space and airborne particle concentration [24–28]. In addition, the effect of vegetation on urban pollutant concentration has often relied on the location of emission sources, especially for the surrounding environment of urban traffic [14,25].

Therefore, different types of green space with different structures in Baoji city, China were selected on the basis of a biotope mapping scheme for investigating the relationships between atmospheric particulate matter concentration and green spaces with different vegetation structures, based on the “matrix effect” of environmental factors, i.e., location, time, wind velocity, temperature, humidity, area to the concentration of PM_{2.5} and PM₁₀ in the green spaces. Hopefully, the results will provide a theoretical basis and practical methods for the optimisation of urban green space structures for improving urban air quality effectively in the future.

2. Materials and Methods

2.1. Study Area

Baoji City is located in the western part of the Guanzhong Plain in Northwest China. The terrain and landscape are complex. The city is surrounded on three sides by mountains, and the Weihe River passes through the centre of the city. It belongs to the semi-humid climate in the warm temperate zone. The average annual temperature is 13.0 °C. The lowest temperature is in January, and the average minimum temperature is 3.5 °C. The highest temperature is in July, and the average maximum temperature is 30.9 °C. The average annual rainfall is about 710–1000 mm [29].

In this study, 11 urban green space types with distinctive differences in vegetation structures were selected, with a distribution along both sides of the Weihe River, which is the largest tributary to the Yellow River in the Baoji City, thereby representing the diversity of green space types, i.e., People’s Park (G1), Weihe Ecological Park (G2), Botanical Garden (G3), Weihe Wetland Park (G4), Baoji University of Arts and Sciences (G5), High-tech Square (G6), Affiliated green space of Panlong Bridge (G7), City Mansion (G8), Ronghai Shengshi (G9), Jufeng Botanical Garden Ecological Community (G10) (Figure 1). The environmental conditions around the ten study areas were similar, and there were no obvious sources of pollutant emission such as factories and boilers, but it was inevitable around the urban

main roads. To avoid the automobile exhaust gas from having a huge impact on the concentration of airborne particulate matter, the selection of specific research plots in the study area was far away from the surrounding area and maintained a certain distance from the peripheral roads. In each study area, a variety of vegetation types and hard-pavement control groups were selected for monitoring and comparison.

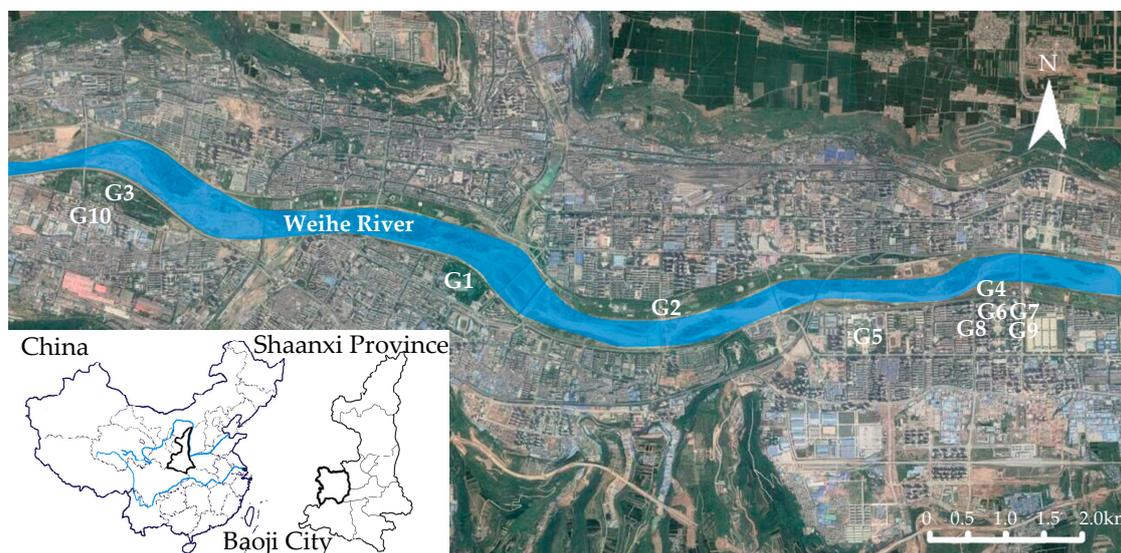


Figure 1. Location of the ten study areas (G1–G10).

2.2. Classification of Green Space Based on Vegetation Structure

In this study, combined with the local characteristics, the green space classification of Baoji was established based on the National Natural Science Fund Project of China “Construction of urban biodiversity conservation system based on biotope mapping”. The classification was modified and classified into three levels according to the vegetation structure [30] (Table 1): the first level was based on the horizontal structure of vegetation according to the canopy cover ratio of trees and shrubs. However, semi-open green space with a 10–30% canopy of trees and shrubs was excluded due to the limited sample; the second level was based on vegetation type including broad-leaved, coniferous and mixed woodland. The third level was mainly focused on the vertical structure, which refers to the vegetation elements at different heights of the tree layer and shrub layer in the vertical form of the combination [31]. In this study, one-layered structures only contained a tree layer, more-than-one-layered structures referred to double tree layers or tree layer and shrub layer combinations.

Table 1. Green space classification of Baoji City based on a biotope mapping [30].

Level	Green Space										
Level 1	Open green space (<10% trees/shrubs)			Semi-closed green space (30%–70% trees/shrubs)				Closed green space (>70% trees/shrubs)			
Level 2	lawn	Shrub	Broadleaved	Coniferous	Mixed	Broadleaved	Coniferous	Mixed			
Level 3	-	-	1-layered	>1-layered		1-layered		>1-layered			
Code	OL ¹	OS ²	S1B ³	S2B ⁴	S1C ⁵	S1M ⁶	S2M ⁷	C1B ⁸	C2B ⁹	C1C ¹⁰	C1M ¹¹

¹ OL: Open green space dominated with lawn; ² OS: Open green space dominated with shrubs; ³ S1B: Semi-closed green space of one-layered broad leaved trees; ⁴ S2B: semi-closed green space of double-layered broad leaved trees; ⁵ S1C: Semi-closed green space of one-layered coniferous trees; ⁶ S1M: Semi-closed green space of one-layered mixed trees; ⁷ S2M: Semi-closed green space of double-layered mixed trees; ⁸ C1B: Closed green space of one-layered broad leaved trees; ⁹ C2B: Closed green space of double-layered broad leaved trees; ¹⁰ C1C: Closed green space of one-layered coniferous trees; ¹¹ C1M: Closed green space of one-layered mixed trees.

2.3. Data Collection

The sample plots were randomly selected for data collection in the ten pre-coded green spaces according to the established classification of vegetation structure. A control group (C) was selected at the hard ground in each of the ten study areas for comparative analysis. The repetition and distribution of each type of green spaces is as follows (Table 2 and Figure 2).

The concentrations of PM2.5 and PM10 and the meteorological factors including wind velocity, temperature and humidity were tested twice in green spaces with typical vegetation structures in each sample plot every two hours, five times a day from 8 a.m. to 6 p.m. in good weather conditions. Monitoring time was from April to May, 2017, as the urban heating period in this quarter has just ended, meaning that the external pollution is reduced. In addition, the monitoring time was when the plants were in the long-leaf stage with a good growth, which can also help reduce the concentration of particulate matter. Data was recorded by a hand-held aerosol mass spectrometer (Metone 831) and a hand-held weather station (FC-36025). All plots were monitored within the same time period and the monitoring sequence was consistent. All the tests were performed at a height of 1.5 m, which is the average height of human respiration. As to the area calculation of each sample plot, it was measured by using a hand-held GPS receiver (Garmin GPS map 629sc) to obtain the latitude and longitude around the sample site and then introducing the plot coordinates into ArcGIS 10.2 software combined with the satellite image to calculate the area accurately.

Table 2. Numbers of 11 different vegetation structure types and hard-pavement control group in 10 study areas.

Vegetation Structures	Number of Sampling Plots in Each Study Area										Total
	G1	G2	G3	G4	G5	G6	G7	G8	G9	G10	
OL		1	1	1	3				2		8
OS		2		1	2	1				1	7
S1B	1	2			1	1	1	2		1	9
S2B		1	1	1	1	1	1	1	1	1	9
S1C	2		1								3
S1M	1	1		1	2		1				6
S2M	2		1		2		1				6
C1B	1	1	1		2		1				6
C2B			3								3
C1C	1	1									2
C1M	3										3
Control Group (C)	1	1	1	1	1	1	1	1	1	1	10
Sum	12	10	9	5	14	4	6	4	4	4	72



Figure 2. Distribution of 11 different vegetation structure plots and their repetition.

2.4. Data Analysis

In this study, Microsoft Office Excel 2007 software (Microsoft Corporation, Redmond, WA, USA) was used for all data recording and collection. Paired samples *t*-test was used to determine whether there were differences in atmospheric particulate matter concentration between the green spaces and control groups. Data were extracted and root transformed to stabilise the variance of individual properties where necessary [32]. Variance analysis was used to determine whether there was a significant relationship between atmospheric particulate matter concentration and green spaces with different vegetation structure, based on the “matrix effect” of environmental factors, e.g., location, time, wind velocity, temperature, humidity and area. Then, generalised regression analysis was applied to analyse the specific effect of a variable on the concentration of PM_{2.5} and PM₁₀ [33]. The statistical analyses were conducted using the statistical software package Minitab 16 (State College, PA, USA). The accepted significance level was at $p < 0.05$.

3. Results

3.1. Effects of Factors on PM Concentration

The area of the selected green spaces ranged from 0.01–1.65 hectares. The local temperature varied from 18.6 to 37.3 °C and the relative humidity range was 25.9–71.9%. The varieties of wind velocity ranged from 0–2.8 m·s⁻¹. The statistical results showed that the location, testing-time, wind velocity, temperature and humidity had highly significant influences on particulate matter concentration ($p = 0.000$), while the area of the site had no significant influence on PM_{2.5} ($p = 0.983$) and PM₁₀ ($p = 0.126$) concentration, and the same for the vegetation structure and PM_{2.5} ($p = 0.500$); however, vegetation structure had a highly significant effect on PM₁₀ concentration ($p = 0.002$) (Table 3).

3.2. Effects of Environmental Factors on PM Concentration

There were significant differences in particulate matter concentrations in 10 different locations (Table 4 and Figure 3). Among them, the generalised regression coefficient of the People’s Park (G1), Weihe River Ecological Park (G2), Botanical Garden (G3) and Jufeng Botanical Garden Ecological Community (G10) were negative, and the other six regional generalised regression coefficients were positive. The reducing effect on the concentration of particulate matter in the study areas G1, G2, G3 and G10 were significantly better than the other six study areas.

The concentrations of particulate matter in the different monitoring time periods were significantly different (Table 5 and Figure 4). From 8:00 a.m. to 6:00 p.m., the concentration showed a downward trend and then gradually increased. It was the lowest in the noon and the highest in the morning.

Meteorological factors such as wind velocity, temperature and humidity had a significant effect on the concentration of particulate matter (Figure 5). In sunny and breezy weather conditions, particulate matter concentration increased with the wind velocity between 0 and 2.8 m·s⁻¹ and humidity between 25.9 and 71.9%, and decreased with the temperature between 18.6 and 37.3 °C. The site area, with a change between 0.01 and 1.65 hectares, had no significant effect on particulate matter concentration (Figure 5).

Table 3. Analysis of Variance of location, time, wind velocity, temperature, humidity, area and vegetation structure influencing PM concentration.

Factors	PM	PM2.5					PM10				
	Df ¹	Seq SS ²	Adj SS ³	Adj MS ⁴	F ⁵	P ⁶	Seq SS ²	Adj SS ³	Adj MS ⁴	F ⁵	P ⁶
Location	9	46.70	267.16	29.685	49.227	0.000	439.18	956.38	106.264	32.321	0.000
Time	4	384.23	31.87	7.967	13.212	0.000	1088.55	1128.94	282.234	85.843	0.000
Wind velocity	1	63.87	63.87	63.868	105.914	0.000	140.70	140.70	140.703	42.795	0.000
Temperature	1	3.84	112.82	112.825	187.098	0.000	63.00	369.94	369.939	112.518	0.000
Humidity	1	514.45	496.89	496.888	823.996	0.000	934.76	899.86	899.863	273.697	0.000
Area	1	0.45	0.00	0.000	0.000	0.983	11.41	7.69	7.692	2.340	0.126
Vegetation structure	10	3.01	5.64	0.564	0.935	0.500	65.25	89.69	8.969	2.728	0.002

¹ Df: Degree of freedom; ² Seq SS: Sequential sum of squares of deviations; ³ Adj SS: Adjusted sum of squares of deviations; ⁴ Adj MS: Adjusted mean square; ⁵ F: Variance test volume;

⁶ P: Significant test of regression equation.

Table 4. Generalised regression analysis of PM concentration and different study areas.

Location	PM	PM2.5				PM10			
	C ¹	SE ²	T ³	P ⁴	C ¹	SE ²	T ³	P ⁴	
G1	−0.816	0.069	−11.860	0.000	−1.465	0.161	−9.119	0.000	
G2	−0.205	0.052	−3.972	0.000	−0.498	0.121	−4.124	0.000	
G3	−0.968	0.075	−12.864	0.000	−1.792	0.176	−10.199	0.000	
G4	0.263	0.072	3.665	0.000	0.329	0.168	1.964	0.050	
G5	0.226	0.051	4.437	0.000	0.220	0.119	1.847	0.065	
G6	0.635	0.083	7.691	0.000	1.471	0.193	7.630	0.000	
G7	0.605	0.065	9.357	0.000	1.334	0.151	8.828	0.000	
G8	0.492	0.085	5.775	0.000	0.683	0.199	3.430	0.001	
G9	0.539	0.090	5.993	0.000	1.021	0.210	4.860	0.000	
G10	−0.772	0.083	−9.283	0.000	−1.302	0.194	−6.710	0.000	

¹ C: Coefficient; ² SE: Coefficient standard error; ³ T: Significant test of regression parameters; ⁴ P: Significant test of regression equation.

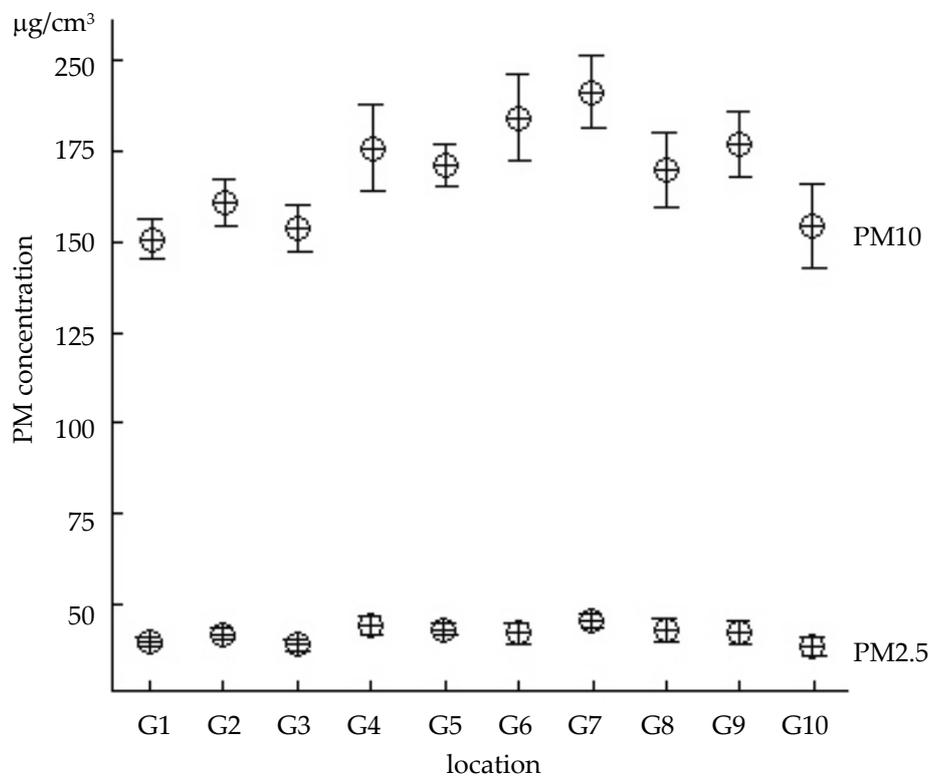


Figure 3. The relationship between location and PM concentration.

Table 5. Generalised regression analysis of PM concentration and different time periods.

Time	PM	PM2.5				PM10			
	C	SE	T	P	C	SE	T	P	
08:00–10:00	0.214	0.078	2.727	0.006	0.442	0.183	2.412	0.016	
10:00–12:00	0.230	0.038	6.017	0.000	0.101	0.089	1.132	0.258	
12:00–14:00	−0.193	0.046	−4.197	0.000	−1.409	0.107	−13.114	0.000	
14:00–16:00	−0.234	0.047	−4.974	0.000	−0.272	0.110	−2.475	0.013	
16:00–18:00	−0.016	0.044	−0.374	0.708	1.139	0.102	11.149	0.000	

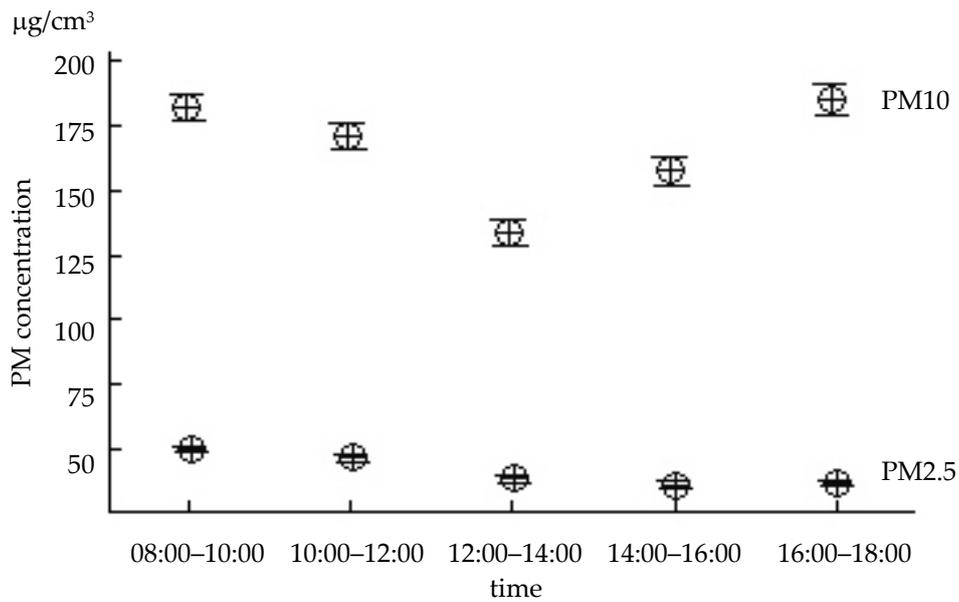


Figure 4. The relationship between time and PM concentration.

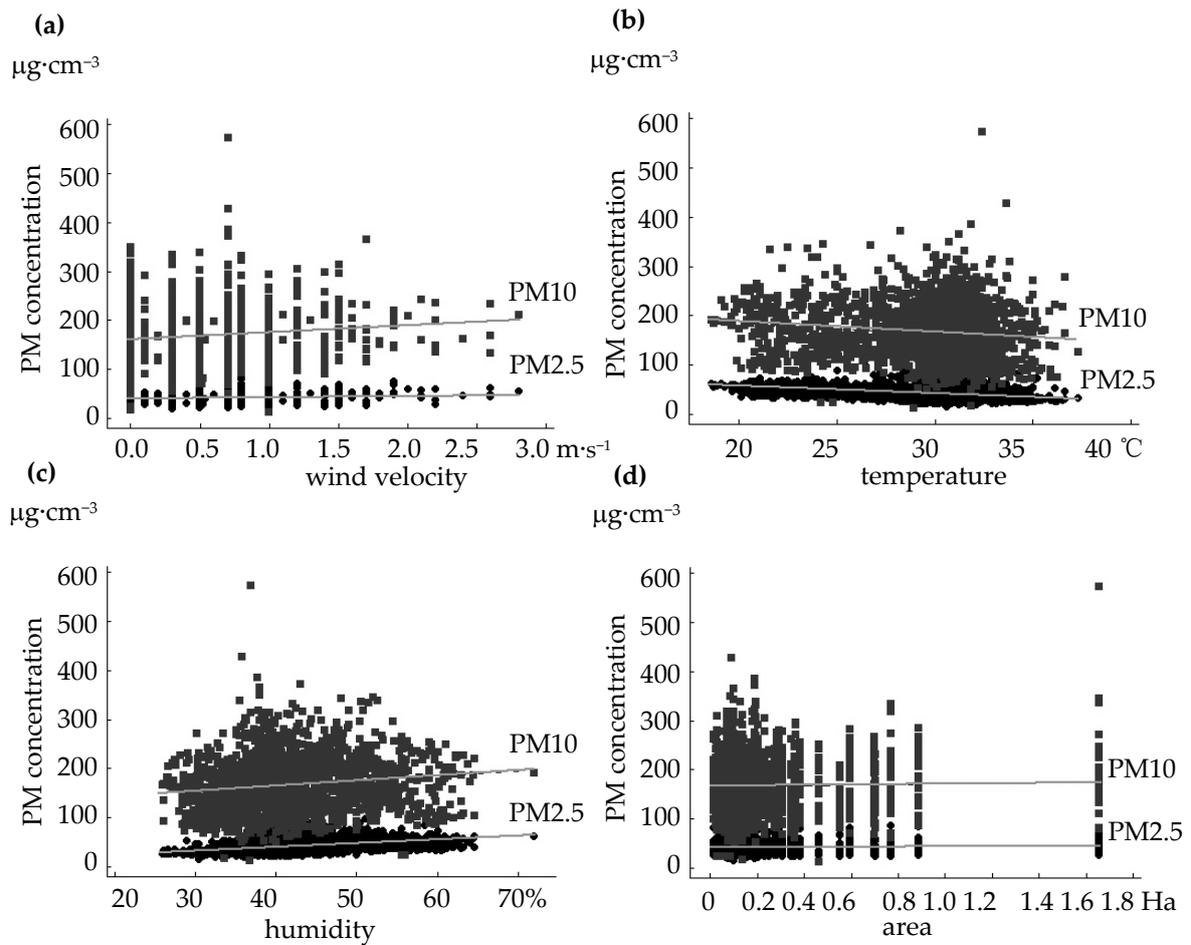


Figure 5. The relationship between wind velocity (a), temperature (b), humidity (c), area (d) and PM concentration.

3.3. Effects of Vegetation Structural Factors on PM Concentration

There were no significant differences between any of the green spaces with different vegetation structures and the control groups after paired samples *t*-test analysis ($p = 0.899$ for PM2.5 and $p = 0.391$ for PM10), indicating that the reducing effect on atmospheric particulate matter concentration by green spaces was very limited under these conditions.

There was no significant difference in the reduction of PM2.5 concentration between different vegetation structural green spaces (Table 6), which could indicate that the role of the 11 vegetation structural green spaces was similar to the reduction of PM2.5 concentration. Meanwhile, there were significant differences in the reduction of PM10 concentration among 11 different vegetation structural green spaces (Table 6). The concentrations of PM10 in the open green spaces with lawns and shrubs (OL and OS) and the double-layered closed broad-leaved forests (C2B) were significantly different from those of the other 9 kinds of vegetation structure in green spaces. The positive and negative coefficients of the generalised regression represent the positive and negative effects of vegetation structure on the concentration of airborne particles. The open green spaces with lawns and shrubs (OL and OS) had a negative effect on PM10 concentration, indicating that the concentrations of PM10 in these types of green space were the lowest. In contrast, the concentration of PM10 was the highest in double-layered closed broad-leaved forests (C2B) due to their positive effect. Meanwhile, the concentrations of PM10 in the other 9 vegetation structures in green spaces were placed in the middle without significant differences.

Table 6. Generalised regression analysis of PM concentration and vegetation structure.

Vegetation Structure	PM	PM2.5				PM10			
	C	SE	T	P	C	SE	T	P	
OL	−0.062	0.059	−1.054	0.292	−0.318	0.138	−2.297	0.022	
OS	−0.115	0.061	−1.884	0.060	−0.444	0.143	−3.106	0.002	
S1B	0.037	0.056	0.669	0.504	−0.144	0.130	−1.105	0.270	
S2B	−0.062	0.056	−1.113	0.266	0.053	0.130	0.403	0.687	
S1C	−0.050	0.084	−0.597	0.551	−0.256	0.196	−1.309	0.191	
S1M	−0.011	0.060	−0.186	0.853	−0.133	0.139	−0.954	0.340	
S2M	−0.002	0.058	−0.034	0.973	0.155	0.136	1.143	0.253	
C1B	0.016	0.058	0.271	0.786	−0.049	0.135	−0.363	0.716	
C2B	−0.012	0.098	−0.121	0.904	0.685	0.228	3.007	0.003	
C1C	0.183	0.098	1.868	0.062	0.246	0.228	1.080	0.280	
C1M	0.079	0.091	0.869	0.385	0.205	0.213	0.959	0.338	

4. Discussion

4.1. Effects of Environmental Factors on PM Concentration

All the study plots and their repetitions in this study were distributed evenly in ten study areas to reduce the interference caused by different environmental conditions in different study areas and their surroundings. In addition, they were located in the city, with similar surrounding environments and no obvious sources of pollutants and were a long distance from motorways; nevertheless, it was still found that the location had a highly significant impact on the airborne particulate concentrations. The concentration of PM2.5 and PM10 in the People’s Park (G1), Botanical Garden (G3) and Jufeng Botanical Garden Ecological Community (G10), which was close to the botanical garden, were the lowest, because the conservation and management of green spaces in these areas was high, ensuring that the plants were regularly trimmed, watered and fertilised. This result was consistent with a previous study which showed that the level of plant management has a certain impact on its ability of reducing air particles [34]. The higher the degree of plant management, the greater the ability to reduce airborne particulate matter. Through regular pruning and watering, the particles would be removed from the plants’ surface restoring their ability for dust retention. Since the traffic flow around the High-tech Square (G6) and affiliated green space of Panlong Bridge (G7)

is higher than other locations, the PM_{2.5} and PM₁₀ concentrations are higher, due to vehicle exhaust emissions. Leonard et al. [14] showed that vehicle exhaust emissions have a significant impact on airborne particulate concentrations, and tight control of relevant measures can effectively reduce airborne particulate concentrations [35].

This study found that the time of monitoring had a highly significant effect on the concentration of airborne particulates, which could be attributed to the interaction between the surrounding pollution sources and the changes in meteorological factors during the day [19,26]. The concentration of airborne particles during the day showed a slight increase after a decreasing trend. It was high in the morning, declined in the afternoon, and gradually rose in the evening; the lowest value of PM₁₀ concentration was found between 12 a.m. and 2 p.m., while the minimum PM_{2.5} concentration was between 2 p.m. and 4 p.m. Huge traffic flow, a large number of automobile exhaust emissions, and serious dust may cause airborne particulate matter concentration to increase in the morning and to reach its peak in the evening. At noon, airborne particulate concentration was low due to there being fewer people and less traffic flow and lighter pollution. Atmospheric particulate matter concentration could be positively correlated with pedestrian flow and traffic flow [36,37].

To avoid the great influence of dramatic changes in meteorological factors on the concentration of airborne particulates, the days with sunny and breeze weather conditions were selected for test. However, the results still showed that meteorological factors had a highly significant effect on the concentration of PM_{2.5} and PM₁₀, and their concentrations increased with wind velocity and humidity, and decreased with temperature. In the same weather conditions, the PM₁₀ concentration range was greater than the range of PM_{2.5} concentration, which could indicate that the larger the particle size of the particles, the more easily transported or settled, and the more obvious the response to meteorological factors [26,37,38].

Wind velocity affects airflow and the diffusion of airborne particulates. Freer-Smith et al. [39] found that the particulate matter sedimentation rate under gale conditions ($9 \text{ m}\cdot\text{s}^{-1}$) was higher than under breeze conditions ($3 \text{ m}\cdot\text{s}^{-1}$). Beckett et al. [40] found that when the wind velocity was less than $8 \text{ m}\cdot\text{s}^{-1}$, the particulate matter sedimentation rates increased with the increase of wind velocity, but the increase of wind velocity may lead to a decrease in the particulate matter sedimentation rate. Wang et al. [27] showed that the concentration of airborne particles increased first and then decreased with the increase of wind velocity, and reached its peak at a maximum wind speed of 14 m/s. Wind speed significantly influenced particulate matter capture efficiency of plants, and particulate matter leaf deposition was positively correlated with low wind speed, but it may decrease with wind speed above a certain critical range [39,40]. The resuspension and deposition rates both varied with different sizes even if the wind speed was similar [41], i.e., large particles and coarse particulates were easily blown off, while fine particulate matter distributed in the grooves of leaf surface was not readily blown away [27,42]. That is, in this study, under breeze conditions, the wind velocity in the range of $3 \text{ m}\cdot\text{s}^{-1}$ was too small to sufficiently diffuse the air particles, resulting in an increase in the concentration. In addition, the wind direction also has a certain influence on the diffusion of air particles. The wind speed studied in this study was low, so this direction was not explored in detail. In future research, wind direction can be increased, especially in the study of airborne particles in traffic-crowded streets.

With the increase of temperature, the convective effect of the atmosphere in the vertical direction was more frequent. This gas circulation exchange accelerated the transportation of air particles, helping reduce the concentration of air particles. Furthermore, with the temperature increased, plant photosynthesis and adsorption of particulate matter were strengthened, helping adsorb airborne particles [26,43,44].

Within a certain range, fine particles were more easily condensed as coagulation nuclei as the humidity increased, resulting in an increase in air particulate concentration. When the relative humidity increased to a certain extent, the amount of wet deposition increased, and then the particle concentration decreased [26,28]. Additionally, an increase in wettability and relative humidity can trigger certain emission mechanisms of biological particles, such as the active wet ejection of

fungus spores or hygroscopic swelling-induced pollen fragmentation, increasing the concentration of atmosphere biological particles around vegetation [45].

In this study, the changes of green space area between 0.01 and 1.65 ha had no significant influence on the concentration of PM_{2.5} or PM₁₀. There were also no significant differences between green spaces and control groups.

That is, the reducing effect on atmospheric particulate matter concentration by green spaces compared to gray spaces was very limited at this area of scale. Green space area was one of the factors that affected the ability of particulate matter dust retention. Within a certain range, the larger the area is, the better the reduction of particulate matter concentration in green space. Liu et al. [46] found that PM_{2.5} concentration index was negatively correlated with forest area. Similarly, the concentration of PM₁₀ in the air was also largely influenced by the green space area factor, increasing the green space area helps to reduce the PM₁₀ concentration in the air [47]. Previous studies have shown that the larger the area of green space, such as forests and parks in the city, the greater the reduction in airborne particulates [48]. Urban areas with proportionally higher concentrations of urban forestry may experience better air quality with regard to reduced ambient particulate matter [24]. In urban land use and land cover planning, increasing the coverage of green land can effectively reduce the concentration of air pollutants [49,50]. All the plots involved in this study were less than 2 ha, and our results confirmed the former. The reduction of particulate matter concentration in the air at this area of scale was not significant, so there was no significant difference in particulate matter dust retention capacity. However, determining the critical value of green area that could reduce the concentration of airborne particulates should be the focus further study. In this study, in which the area of green space was no more than 2 ha, the change in area was not sufficient to produce qualitative changes in the concentration of particulate matter in the air. This would be an important implication for the purposes of eco-oriented design of green space planning and design, in that the configuration of different plants at such a small scale could be very limited in terms of purifying the air particulate matter.

4.2. Effects of Vegetation Structural Factors on PM Concentration

There was no significant difference in the reduction of PM_{2.5} concentration in different vegetation structures in green spaces, while there were significant differences in PM₁₀ concentration. Plants use their special micro-morphological structure to retain particulate matter. The ratio of the mass and volume of PM₁₀ on the leaf surface was much larger than that of PM_{2.5} [51]; that is, the plant had a stronger reduction effect on the larger particle size, which may cause this difference.

The concentration of PM₁₀ in the open green spaces dominated, with lawns and shrubs being the lowest, while that of double-layered closed broad-leaved forests was the highest. Therefore, green spaces with different vegetation structures can be scientifically chosen for adsorption of particulate matter concentration in air quality-oriented green space planning and design. Some researchers have considered that the concentration of airborne particulate matter in the closed plant communities is higher than that in open plant communities [21,52]. Particulate matter leaf depositions on trees and shrubs with complex structures are higher than those of herbs and liana species [53]. Through the analysis of the vegetation characteristics of different vegetation structures, we found that in double-layered closed broad-leaved forest the structure of the vegetation was complex, causing air turbulence, and the plant species were diverse and varied in quantity, hindering the sedimentation of PM₁₀ [42]. The particles adsorbed on the surface of the broadleaf leaves were only temporarily trapped, prone to bounce back, and were then suspended in the atmosphere, thereby increasing the concentration of air particles [54,55]. In the open green spaces dominated by lawns and shrubs, the sedimentation of PM₁₀ was less obstructed, and the settling process could be completed directly; therefore, the concentration of PM₁₀ in the air was low. Another possible reason was that meteorological factors, especially wind, have a greater influence on the concentration of particulate matter in open vegetation structures, and it was possible to transport the pollutant to a new location, reducing the concentration.

In addition to considering the impact of vegetation structure on airborne particulate matter, tree species therein also have a significant impact on air quality. Trees are a source of air pollutants, and they can produce pollen, spores, some bio-volatile organic compounds, and other biological particles, and increase the concentration of air particles after the chemical reaction with other pollutants in the air and the formation of organic aerosols [56–58]. In addition, the particles produced by the plant itself could limit the spread of contaminants and thus increase the concentration of local particulates [59,60]. Therefore, it could also be one of the reasons for increasing the concentration of particulate matter that the large number of plants and their complex layers in double-layered closed broad-leaved forests. In future studies, it is necessary to carry out more detailed classification and to examine whether the difference of specific plant species is significant for the reduction of airborne particle concentration.

5. Conclusions

Based on the detailed classification of the urban green space vegetation structure in Baoji City, combined with meteorological factors and other factors to analyse the PM_{2.5} and PM₁₀ concentrations, we found that the reduction of PM_{2.5} and PM₁₀ concentrations of different vegetation structures in urban green spaces have certain differences and are subject to a variety of common constraint factors.

These findings suggest that if controlling pollution sources cannot be entirely relied upon to control air pollution, changing the horizontal and vertical structure of vegetation and vegetation type can contribute to the reduction of PM_{2.5} and PM₁₀ concentrations to a certain extent. The different combinations of vegetation structure not only consider the plant's own use of some special structures to block the particulate matter, but also consider its impact on meteorological factors, and thus the concentration of particulate matter. According to the different main functions of urban green space in different regions, different vegetation structures can be selected for planting. However, the reduction effect of urban green space less than 2 ha on the concentration of airborne particles was limited and the effect of green space on a larger area scale is more significant. Changing the fragmentation status of urban green space and increasing the area of green space will help better play its ecological benefits. The above results can be used to provide a theoretical basis and practical methods for the optimisation of urban green space structures for improving urban air quality effectively in the future.

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