

# A Climatic Index for Aeolian Desertification in Northern China and its Application to Dust Storm Frequency

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# ABSTRACT

Aeolian desertification is the dominant desertification type in northern China. The robust relations between climate variables and desertification indicators must be established while predicting the future desertification trends by using climate models. Vegetations are sensitive to climate change and desertification simultaneously. A basic feature of aeolian desertification process in northern China is that aeolian sand and dust activities proceed the vegetation growth season in the same year. In this study, such a lagging effect of desertification to climate change is described by the wind erosion and vegetation growth models. A non-dimensional aeolian desertification index composed of wind speed, precipitation and temperature is constructed. Its applicability is confirmed by the temporal trends of dust storm frequency in the dust source regions.

Keywords: Aeolian desertification; Climate variables; Dust storm frequency; Northern China

# INTRODUCTION

Desertification is land degradation at arid, semiarid and dry subhumid areas resulting from various factors, including climatic variations and human activities UNCCD (1994). The effects of future climate change on desertification are of general interest [1,2]. A number of robust models can predict climate variables such as precipitation and temperature under different future emission scenarios [3,4]. Desertification can be represented by many indicators including ground surface albedo, soil organic matter and grain size distribution, vegetation cover, dust emission flux and dust storm frequency etc. [5]. However, the relations between climate variables and desertification indicators have not been well established, although the ratio of precipitation to potential evapotranspiration was used to define meteorological drought and assess desertification [6-8].

Desertification is a serious threat to China where aeolian desertification, water erosion, salinization, and freeze-thaw desertification take place frequently. Aeolian desertification is the dominant desertification type in northern China [9,10]. Since the formation and evolution of dune fields are closely related to climatic variables, desertification trends were previously estimated by the dune mobility index [11], in which the annual duration of aeolian

sand transport was taken into account. Obviously, desertification does not always occur in dunefields. The entrainment, transport and deposition of sand and dust by wind are often associated with the desertification of farmlands and steppes. Specially dust storm directly reflects the dynamical process of severe desertification. A statistical analysis of the data recorded by 129 meteorological stations in northern China during 1960-2007 showed that the predominant variables affecting dust storm frequency are wind speed, temperature and precipitation [12,13]. However, the detailed mechanisms behind the correlation relations between climatic variables and dust storm frequency remain unclear. In this study, a non-dimensional climatic index for aeolian desertification is found using these three variables, and its application to dust storm frequency is displayed.

## MATERIALS AND METHODS

## Index for aeolian desertification

The modes of particle motion in aeolian activities are traditionally classified as suspension, saltation and creep. Most of the mass and energy are transported by saltation. A widely used saltation formula is Lettau and Lettau (1978) [14],

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$$q = \begin{cases} 0 & \text{if } u_* \le u_{*t} \\ c_0 \frac{\rho}{g} \sqrt{\frac{d}{D}} u_*^3 (1 - \frac{u_{*t}}{u_*}) & \text{if } u_* > u_{*t} \end{cases}$$
(1)

where q is saltation flux,  $c_0$  and D are two constants,  $\rho$ , g, d,  $u_*$ ,  $u_{*t}$  are air density, gravity acceleration, sand grain size, friction speed, and threshold friction speed, respectively. The timescales of saltation events are hourly or daily. Whereas climate changes in the timescale of month or year at least. Given a bare, loose and dry land, the annual saltation flux per unit width can be written as,

$$Q_0 = \int_0^{T_0} q dt \tag{2}$$

Where  $T_{\scriptscriptstyle 0}$  is equal to one year, the subscript of "0" denotes the current year.

From Equation (1), Equation (2), the wind speed effect of the i-th year is extracted,

$$Q_0 \propto U_i = \int_0^{T_i} U^3 (1 - \frac{U_i}{U}) dt$$
 (3)

Where U (>Ut) is the 10-meter wind speed measured by the meteorological station, Ut is the corresponding threshold wind speed. Ui is analogous to drift potential [15] Equation (3). It decreases with the averaging time of wind speed [16,17] because saltation is highly intermittent [10,18]. In the following calculation, the averaging time of wind speed and the threshold wind speed is set as 24 h and 4.5 m/s, respectively.

For a vegetated land, the annual saltation flux is,

$$Q_0 = Q_0 \left( 1 - \frac{C}{C_*} \right) \tag{4}$$

Where C is vegetation coverage,  $C_*$  is a threshold above which there is no saltation [19] Equation (4).

We assume that vegetation coverage C is proportional to the aboveground biomass  $M_0$ ,

$$c \propto M_0 = M_0^a + M_0^p \tag{5}$$

Where  $M_0^a$  and  $M_0^p$  are the above-ground biomasses of annuals and perennials Equation (5). There is a lag time between aeolian desertification and vegetation growth in northern China. Aeolian sand and dust activities often occur from March to May. The growing season of vegetation in summer. The dead annuals of last year and non-green perennials protect the current land surface against wind erosion. This fact can be expressed as,

$$M_0^a \propto R_{-1}^a T_{-1} \tag{6a}$$

$$M_0^p = \alpha M_{-1}^p + \alpha R_{-1}^p T_{-1} = \alpha^n M_{-n}^p + \sum_{i=1}^n \alpha^i R_{-i}^p T_{-i}$$
(6b)

where  $\alpha$  is a positive constant less than 1,  $R_p^i$  and  $R_p^i$  are the annual growth rates of annuals and perennials in the i-th year. There are many functional forms for modeling plant growth [20]. For simplicity, the linear growth model is selected in Equation (6).

From Equation (5) and (6), we have,

$$M_{0} \approx \alpha R_{-1}^{a} T_{-1} + \sum_{i=1}^{n} \alpha^{i} R_{-i}^{p} T_{-i}$$
<sup>(7)</sup>

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According to the expression of net primary production rate [21], the annual growth rates can be described by several climatic variables Equation (7),

$$R_i^a = C_a \sum_{j=1}^{12} \frac{p_j^a}{p_0} \exp\left[-\left(\frac{\overline{T}_j - T_{opt}}{T_{opt}}\right)^2\right]$$
(8*a*)

$$R_{i}^{p} = C_{p} \sum_{j=1}^{12} \frac{p_{j}^{p}}{p_{0}} \exp\left[-\left(\frac{\overline{T}_{j} - T_{opt}}{T_{opt}}\right)^{2}\right]$$
(8b)

where Ca and Cp are two constants,  $p_j^a$  and  $p_j^p$  are the daily summed precipitations for the growths of annuals and perennials in the j-th month,  $p_0$  is a characteristic precipitation for non-dimenionalizing,  $\overline{T}_j$  is the monthly mean temperature,  $T_{opt}$  is the optimal temperature for plant growth. The expressions of  $R_a^i$  and  $R_p^i$  in Equation (8) are simplified from the regional vegetation model developed by [21]. For the desert steppes in north-west China, perennial shrubs are dominant plants responsible for surface stabilization. The sap flow measurements revealed that rainfall events larger than a threshold pc can effectively influence the growth of desert shrubs [22]. Hence the precipitations of  $p_i^p$  and  $p_i^a$  can be computed as follows,

$$p_i^p = \begin{cases} 0 & \text{if } p \le p_c \\ p - p_c & \text{if } p > p_c \end{cases}$$

$$\tag{9a}$$

$$p_i^a = \begin{cases} p & \text{if } p \le p_c \\ p_c & \text{if } p > p_c \end{cases}$$
(9b)

where  $p_c=5 \text{ mm}$ , p is the rainfall of one event. For the typical steppes in north-east China, the dominant plants are perennial herbs. So it is assumed that  $p_j^p = p$  and  $p_j^a = 0$  there. The other parameters in Equation (9) are  $C_a = C_p = 1$ ,  $p_0 = 50 \text{ mm}$ ,  $T_{opt} = 20^{\circ}\text{C}$ . The effects of temperature and precipitation in that month are neglected if  $\overline{T_j} < 5^{\circ}\text{C}$ .

$$\overline{U}_i = \frac{U_i - U_i \,|\, \min}{U_i \,|\, \max - U_i \,|\, \min} \tag{10a}$$

$$\bar{M}_{0} = \frac{M_{0} - M_{0} |\min}{M_{0} |\max - M_{0} |\min}$$
(10b)

 $U_i$  and  $M_0$  are normalized by their extremes, where "max" and "min" represent the maximum and minimum values.

 $\overline{M_o}$  quantifies the effects of precipitation and temperature Equation (10). Based upon Equation (3) Equation (5), a climatic index for aeolian desertification can be constructed,

$$I = \overline{U}_i \left( 1 - \overline{M}_0 \right) \tag{11}$$

## RESULTS

#### Application to dust storm frequency

The dataset of dust storm time series in northern China was established about ten years ago, based on a combining criterion of visibility, wind speed, and the weather process [23,24]. It offers high-quality data for many subsequent studies [25-27]. The dust storm dataset in 1954-2007 and the daily surface climate data were all downloaded from the China Meteorological Data Sharing Service System. Here the climatic index of Equation (11) is applied to analyze 17 dust storm frequency (number/year) measured in the dust sources regions; see (Figure 1) for the locations of meteorological stations. These stations are roughly uniformly

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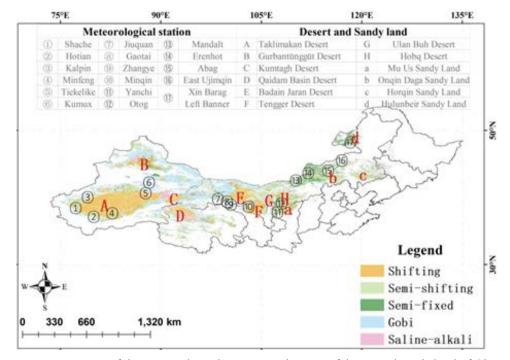


Figure 1: Locations of the meteorological stations in the map of desert and sandy land of China.

distributed in the arid and semi-arid areas of northern China.

The time series can be decomposed into the trend and fluctuation components,

$$I = I_0 + I' \tag{12a}$$

$$N = N_0 + N' \tag{12b}$$

where N is dust storm frequency,  $I_0$  and  $N_0$  are trends, I' and N' are fluctuations. Such decompositions are easily performed by the complete ensemble empirical mode decomposition adaptive noise code available at http://perso.ens-lyon.fr/patrick.flandrin/emd. html.  $N_0$  should synchronously change with  $I_0$  if the climatic index I is applicable. A simple relation is linear,

$$N_0 = aI_0 + b \tag{13}$$

where a and b are two fitting parameters. The subset of dust storm frequency during 1980-2005 is investigated because the measurement standard of dust visibility was revised in 1980 (Table 1).

 Table 1: Parameters in the linear fitting between the trends of dust storm frequency and climatic index.

Meteo- rological Station		Location	α	n	a	b	$\mathbb{R}^2$
1	Shache	(77°16′E,38°26′N)	0.9	18	0.82	0.153	0.841
2	Hotian	(79°56′E,37°8′N)	0.98	16	1.89	0.06	0.798
3	Kalpin	(79°3′E,40°30′N)	0.77	7	1.31	0.127	0.977
4	Minfeng	(82°43′E,37°4′N)	0.97	16	1.02	0.103	0.87
5	Tiekelike	(87°42′E,40°38′N)	0.93	2	0.63	0.3	0.883
6	Kumux	(88°13′E,42°14′N)	0.97	16	0.84	0.084	0.978
7	Jiuquan	(98°29′E,39°46′N)	0.97	2	1.77	-0.106	0.772
8	Gaotai	(99°50′E,39°22′N)	0.48	1	1.02	0.016	0.95
9	Zhangye	(100°26′E,38°56′N)	0.95	2	1.16	0.013	0.697
10	Minqin	(103°5′E,38°38′N)	0.98	3	1.83	-0.148	0.983

11	Yanchi	(107°24′E,37°47′N)	0.94	5	1.48	-0.055	0.922
12	Otog	(107°59′E,39°6′N)	0.98	1	0.48	0.202	0.991
13	Mandalt	(110°8′E,42°32′N)	0.85	14	4.57	-0.883	0.936
14	Erenhot	(111°58′E,43°39′N)	0.93	10	0.38	0.278	0.449
15	Abag	(114°57′E,44°1′N)	0.85	4	0.34	0.316	0.258
16	East Ujimqin	(116°58′E,45°31′N)	0.81	2	0.55	0.114	0.722
17	Xin Barag Left Banner	(118°16′E,48°13′N)	0.97	8	0.88	0.02	0.977

1 lists all fitting parameters including  $\alpha$  and n in Equation (7) and a and b in Equation (13). R, is the goodness of fit. The positive values of support the hypothesis that N<sub>0</sub> and I<sub>0</sub> vary synchronously. The lags of aeolian desertification to climate change range from 1 to 18 years. Three modern dust source regions are the Taklimakan Desert, the Hexi Corridor and western Inner Mongolia Plateau, and the central Inner Mongolia Plateau [25]. The meteorological stations of Nos. 1-6, 7-10, and 11-17 in Figure 1 are set up in these three regions, respectively. The landscapes of Nos. 13-17 are typical steppes. The changes of dust storm 145 frequency N, climatic index I and their tends N<sub>0</sub> and I<sub>0</sub> with time of three sites, Minfeng, Zhangye, and Erenhot namely, are plotted in Figure 2. Although the detailed temporal curve shapes of N and I do not identical, their tends of No and Io are similar to each other. This demonstrates that the climatic index of Equation (11) is reasonable and thus suitable for prescribing the desertification in northern China. It should be pointed out that the climatic index is derived from the fundamental principles of aeolian and vegetation dynamics. In the current study, only the crucial factors are focused on. In addition to wind speed, precipitation and temperature, other variables such as radiation, evapotranspiration and soil water have influences on desertification. The more advanced climatic indexes should embody them in the future.

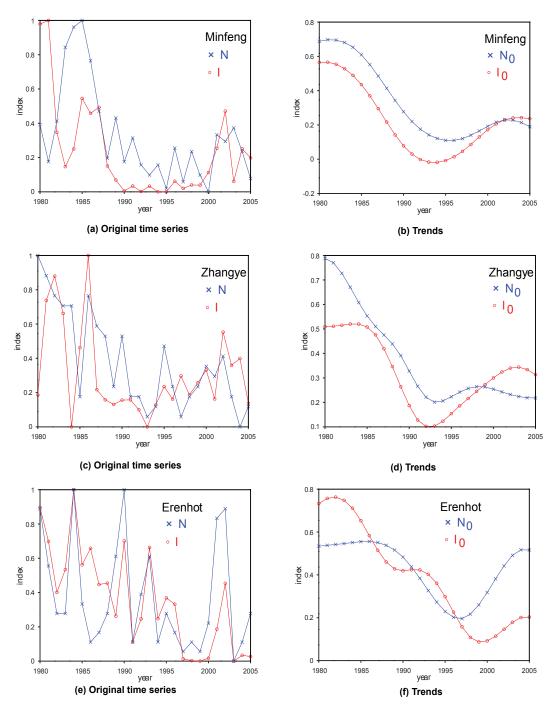


Figure 2: Changes of dust storm frequency N, climatic index I and their tends No and Io with time.

## CONCLUSION

Aeolian desertification is the dominant desertification type in northern China. A basic fact is that aeolian sand and dust activities proceed the vegetation growing season in the same year. Consequently, the surface is generally protected to resist wind erosion by the residual vegetation cover formed in previous years. In the presented study, this phenomenon is quantitatively described by the annual saltation flux under the influence of vegetation cover. The effects of precipitation and temperature are introduced by the linear growth model of the plant. Combing with wind speed, a non-dimensional climatic index for aeolian desertification is established. The trend analysis of dust storms frequent confirms that this climatic index is practical.

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#### Authors' contributions

All authors contributed equally.

#### Data availability

The datasets used and/or analyzed during the current study are available from the corresponding author on reasonable request.

#### **Conflicts of Interest**

The authors declare no conflicts of interest.

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