
A Prediction Model for the Animal Plague in *Spermophilus dauricus* Focus in China

Xiaolei Zhou¹, Boyu Zhang², Xianbin Cong^{1, *}, Xiaoheng Yao¹, Cheng Ju¹, Zhonglai Li², Cheng Xu¹, Tianyi Duan¹, Guijun Zhang¹, Lei Chen¹, Zhencai Liu¹

¹Chinese Base for Control of Plague and Brucellosis, Chinese Center for Disease Control and Prevention, Baicheng, China

²Laboratory of Mathematics and Complex Systems, Ministry of Education, School of Mathematical Sciences, Beijing Normal University, Beijing, China

Email address:

cx0805@sina.com (Xianbin Cong)

To cite this article:

Xiaolei Zhou, Boyu Zhang, Xianbin Cong, Xiaoheng Yao, Cheng Ju, Zhonglai Li, Cheng Xu, Tianyi Duan, Guijun Zhang, Lei Chen, Zhencai Liu. A Prediction Model for the Animal Plague in *Spermophilus dauricus* Focus in China. *Science Journal of Public Health*.

Vol. 3, No. 5, 2015, pp. 612-617. doi: 10.11648/j.sjph.20150305.13

Abstract: Plague is a fatal infectious disease that causes serious harm to humans. Its occurrence threatens not only public life, but also economic development. Although the incidence of plague in China shows a downward trend, the risk of animal and human plague still persists. By analyzing the data of the *Spermophilus dauricus* focus in the Inner Mongolia Autonomous Region from 1981 to 2012, we established a statistical model to predict the epidemic of the animal plague, which combines the best subset regression method and the exponential smoothing method. According to the data from 1981 to 2011, the model predicted that there is no animal plague epidemic risk in 2012. This result is consistent with the report from the Inner Mongolia Autonomous Region, the plague bacillus *Yersinia pestis* was not detected in the *S. dauricus* focus in 2012. In addition, our model can be extended to predict the epidemic of plague in other foci. Potential and limitations of the model are discussed.

Keywords: *Spermophilus dauricus* Focus, Exponential Smoothing Method, Best Regression Subsets Method, Risk Classification

1. Introduction

Plague is a natural epidemic disease, primarily occurring in rodents. It spreads fast, has a high mortality rate, and can cause a human epidemic. The sources of infection are mainly rodents and humans (pneumonic plague), and the infectious vectors are mainly fleas [1]. Several plague epidemics have occurred in China, these epidemics have had a perennial course and a variety of foci covering large areas [2, 3]. According to the geographical landscape, host, insect vectors, and ecological characteristics of the plague bacillus *Yersinia pestis*, there are a total of 12 types of natural foci of plague in China: *Spermophilus dauricus* focus, *Meriones unguiculatus* focus, *Marmota himalayana* focus, *Marmota caudata* focus, *Marmota baibacina*-*Spermophilus undulatus* focus, *Spermophilus alaschanicus* focus, *Microtus brandti* focus, *Marmota sibirica* focus, *Apodemus chevrieri*-*Eothenomys miletus* focus, *Rattus flavipectus* focus, *Microtus fuscus* focus, and *Rhombomys opimus* focus [4, 5]. In the 305 years before the People Republic of China was founded, epidemic plagues

occurred about 179 times in 20 provinces (autonomous regions) including 549 counties (cities, banners), infected 2,598,794 people, and killed 2,399,400 people. From 1950 to 1954, 6,868 cases of human plague were recorded. Since 1955, the outbreak of epidemics of human plague was under control [6]. However, from 1990 to 2002, the number of human infected cases increased due to recrudescence of the animal plague in the *R. flavipectus* foci and its spread to humans. After 2003, the incidences of human plague in China showed a downward trend, with the decline of the *R. flavipectus* plague epidemic.

Because of the high infectivity and fatality, a great deal of research has been devoted to analyze the main factors that affect the epidemic of plague and the change trends of the host populations (in China). Most of these investigations focused on the occurrences of animal plague. For instance, by applying grey system method, Li and Baiyin [7] predicted that the density of *M. sibirica* in Inner Mongolia Autonomous Region

would decrease according to the collected data of 1981-1992. Based on the data from 1981 to 1993 on the *Citellus alaschanicus* plague focus and using the multiple stepwise regression method, Qin and Li [8] showed that the density of *C. alaschanicus* and nest flea index were the main factors affecting the epidemic of plague. Later, Li and Zhang [9] indicated that a combination of the grey system method and the runs testing method can be used to predict the epidemic years in YiKeZhaoMeng (data from 1965 to 1991) and Otog QianQi (data from 1967 to 1990). They further analyzed the data of the *M. brandti* plague focus from 1980 to 1996 by the runs testing method and found epidemic cycles [10]. Using the astronomical, meteorological, and monitored data of the northern desert steppe region in the Inner Mongolia Autonomous Region from 1982 to 1993, Mi et al [11] calculated the epidemic intensity by the multiple regression and optimal regression models. In the next few years, multiple regression, fuzzy clustering analysis, discriminant analysis, and grey system methods were applied to find out the main influencing factors of the *S. dauricus* plague focus based on part of the monitored and meteorological data from 1957 to 2000. All these methods showed that the epidemic of plague is mainly affected by rodent density, body flea index, and burrow-dwelling flea index [12, 13, 14, 15]. Recently, prediction models based on the geographic information system have attracted considerable attentions [16], and these methods have been applied to *S. dauricus* focus, regions along the Qinghai-Tibet Railway and Guangxi Province [17, 18, 19]. In contrast to animal plague, few studies considered human plague in China since the infection of human plague is very rarely accident in the recent years. Zhang et al. [20] linked human plague intensity from 1871 to 2003 to climate at the province scale. Using the spatial and temporal human plague records in China from 1850 to 1964, Xu et al. [21] revealed that human plague intensity in northern or southern China is positively related to wetness index. They further showed that wet climate and transportation routes accelerate spread of human plague [22].

The purpose of our study is to predict the occurrence of animal plague epidemic in the *S. dauricus* focus. The earliest written record of the *S. dauricus* focus was Wu Lien-teh's report of 1917-1920 on the "Tongliao bean rat plague (*S. dauricus* plague)" [23]. Subsequently, *Y. pestis* was isolated from *Rattus norvegicus*, *Mus musculus*, *Mesocricetus auratus*, and *Apodemus agrarius* in 1937. This was the first bacteriological evidence confirmed the existence of this focus. In 1957, researchers from the former Epidemiology Institute of Chinese Academy of Medical Sciences, Jilin province, and the Inner Mongolia Autonomous Region indicated that *S. dauricus* was the main host of the focus and *R. norvegicus* was the second [24]. Although the incidence of plague has declined significantly in recent years, the risk of animal and human plague epidemic still persists. Between 1981 and 2012, a total of 87 strains of *Y. pestis* were detected, where 64 strains

were from animals (1985, 1986, 1987, 1988, 1990, 1994, 1996 and 1998) and 23 strains were from vectors (1985, 1987, 1988, 1989, 1990, 1996 and 1997) (Section 2.1.2 Date source). It is important to note that more than 95% of the strains were detected in the Inner Mongolia Autonomous Region (83 out of 87), and the other 4 strains were from Jinlin Province (1985). In addition, *Y. pestis* was not detected in Liaoning province and Heilongjiang province in the last 30 years.

In this study, we establish a prediction model based on the best subset regression (BRS) method and the exponential smoothing method. The BRS method is a variable selection method that attempts to the "best" subset of a set of predictor variables in regression [25]. This method has been applied to analyze the main influencing factors of plague [26, 27]. The exponential smoothing method is a time series analysis model developed on the basis of a moving average method [28]. It emphasizes that the development of a time series is mainly affected by recent data, where the predicting value (called the exponential smoothing value) is the weighted average of the observation value and the exponential smoothing value of the previous period. Comparing with other moving average methods (e.g., ARMA, ARIMA), the exponential smoothing method only requires one latest observation value and one latest predicting value, therefore can greatly reduce the data storage problem. The exponential smoothing method has been effectively used to predict the frequency changes of ecological variables, such as population density and biomass of rodent communities [29]. Therefore, we could expect that it has a good potential for predicting animal plagues, as we will show later.

2. Materials and Methods

2.1. Materials

2.1.1. The *S. Dauricus* Focus

S. dauricus focus are located between north latitude 41°31' to 46°45' and east longitude 113° to 126°41' [30]. The specific location of the focus is at the south of the Lalin River in Jilin Province, west of the Yitong and Kacha Rivers, east of the Mongolian desert plateau, and north of the Yinshan Mountains (south to the Nuluer Tiger Mountain region of Liaoning Province), where includes the Liaohe River basin and the steppe zone from the southern foothill of the Greater Khingan Mountain to the south of the Otindag sandy land [30]. *S. dauricus* focus is distributed in the 53 counties (cities, banners) of Inner Mongolia Autonomous Region, Jilin province, Liaoning province, and Heilongjiang province, covering an area of 161,918 km² (see Figure 1, yellow and blue regions). Since *Y. pestis* was rarely detected in Jilin province, Liaoning province, and Heilongjiang province, we only focus on the *S. dauricus* focus in the Inner Mongolia Autonomous Region (see Figure 1, blue region).

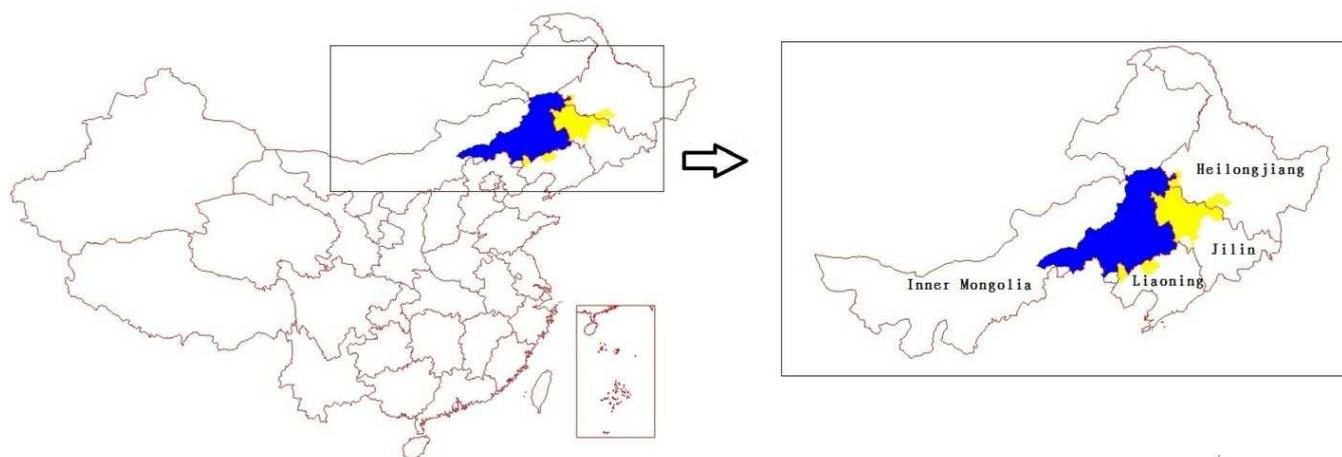


Figure 1. Distribution of the *S. dauricus* focus (yellow and blue regions). The *S. dauricus* focus is mainly distributed in the Inner Mongolia Autonomous Region (blue region).

2.1.2. Data Source

The meteorological, geographical and monitored data of the *S. dauricus* focus in the Inner Mongolia Autonomous Region from 1981 to 2012 are used for the animal plague prediction. To be precise, monitored data includes 7 factors: density of *S. dauricus*, *S. dauricus* flea infection rate, *S. dauricus* flea index, nest flea infection rate, nest flea index, burrow-dwelling flea infection rate and burrow-dwelling flea index; meteorological data includes 7 factors: temperature, relative humidity, precipitation, air pressure, relative temperature, evaporation, and sunshine; geographical data includes 3 factors: living environment, vegetation and soil. Data are collected from annual summary of the Inner Mongolia Autonomous Region

from and the mid-term evaluation report of the “Eleventh Five-Year plan”.

2.2. Methods

2.2.1. A Basic Framework of the Prediction Model

Our prediction model consists of four phases. First, main factors that affect the epidemic are selected using the BRS method. Second, a risk classification model is established to evaluate the risk of epidemic based on the fitting results of the BRS method. Third, the exponential smoothing method is applied to predict the main factors in the next period. Finally, the risk of epidemic in the next period is calculated according to the predicted results and the best regression equations.

Table 1. The best regression equations with different numbers of factors.

Numbers of factors	Best regression equations	R ²
1	$y = -0.097 + 0.152X_3$	0.196
2	$y = -0.428 + 0.134X_3 + 0.977X_7$	0.298
3	$y = 0.309 + 0.384X_1 - 0.022X_2 + 0.248X_3$	0.390
4	$y = 1.032 + 0.556X_1 - 0.026X_2 + 0.266X_3 - 0.012X_4$	0.473
5	$y = 0.948 + 0.515X_1 - 0.028X_2 + 0.267X_3 - 0.011X_4 + 0.011X_6$	0.486
6	$y = 0.963 + 0.559X_1 - 0.029X_2 + 0.279X_3 - 0.010X_4 - 0.009X_5 + 0.011X_6$	0.491
7	$y = 0.890 + 0.548X_1 - 0.027X_2 + 0.270X_3 - 0.010X_4 - 0.009X_5 + 0.008X_6 + 0.167X_7$	0.493

Note: X_1 – X_7 represent the density of *S. dauricus*, *S. dauricus* flea infection rate, *S. dauricus* flea index, nest flea infection rate, nest flea index, burrow-dwelling flea infection rate, and burrow-dwelling flea index, respectively. R^2 is the coefficient of total correlation. Statistical software: Matlab.

2.2.2. Best Regression Equations

Preliminary analysis based on the BRS method showed that the *S. dauricus* plague epidemic was mainly decided by the characteristics of the *S. dauricus* population itself, such as the density of *S. dauricus*, and was less affected by the climate and environment [27]. Therefore, we only consider the following 7 factors as explanatory variables when establishing the best regression equations: density of *S. dauricus*, X_1 ; *S. dauricus* flea infection rate, X_2 ; *S. dauricus* flea index, X_3 ; nest flea infection rate, X_4 ; nest flea index, X_5 ; burrow-dwelling flea infection rate, X_6 ; and burrow-dwelling flea index, X_7 . In

the model, the dependent variable $y = 1$ if *Y. pestis* was detected, and $y = 0$ if no bacterium detected. The best regression equations with different numbers of factors are shown in Table 1. In particular, the four main factors of the *S. dauricus* focus screened out by the BSR method are the density of *S. dauricus*, *S. dauricus* flea infection rate, *S. dauricus* flea index, and the nest dye flea rate.

2.2.3. Risk Classification

In a previous study of the risk classification of *S. dauricus* focus [31], there are 3 different risk levels according to the value of the dependent variable (of the BRS method): If $y > 2/3$,

then the focus is predicted as an epidemic. If $y < 1/3$, then the focus is predicted as no epidemic risk. If $1/3 \leq y \leq 2/3$, then the focus is predicted as a high-risk area. According to the collected data from 1981 to 2011, when $y > 2/3$, the accuracy of epidemic prediction was 100%. When the number of factors in the best regression equations is greater than 3 and $y < 1/3$, the accuracy of non-epidemic prediction was 100%. When $1/3 \leq y \leq 2/3$, the fitting rate of epidemic prediction was approximately 50% (see Table 2). In this study, we adopt this risk classification standard.

Table 2. Accuracy of the prediction for different numbers of factors based on the risk classification standard.

Numbers of factors	Epidemic (%)	No-risk (%)	High-risk (%)
1	100	77	20
2	100	82	40
3	100	89	44
4	100	100	45
5	100	100	42
6	100	100	56
7	100	100	56

2.2.4. Exponential Smoothing

The basic formula of the exponential smoothing method is

$$ES(t) = a \times X(t-1) + (1-a) \times ES(t-1),$$

where $ES(t)$ represents the predicting value (i.e., the exponential smoothing value) at time t , where the range of t is from 1982 to 2012; $X(t)$ represents the observation value at time t ; and a is smoothing constant in the range $[0,1]$. We extrapolate the trends of the 7 main factors (X_1-X_7) by using the exponential smoothing method. Since monitored data in 1981 are incomplete, we cannot calculate $ES(1982)$ from the formula. Instead, we assign the observation value to be the initial smoothing value, i.e., $ES(1982)=X(1982)$. To determine the smoothing constant a , we use the minimum mean square error (MMSE) principle. According to this principle, the optimal smoothing constants of the 7 main factors are as follows: (1) density of *S. dauricus*: $a_1 = 0.81$; (2) *S. dauricus* flea infection rate: $a_2 = 0.45$; (3) *S. dauricus* flea index: $a_3 = 0.44$; (4) nest flea infection rate: $a_4 = 0.94$; (5) nest flea index: $a_5 = 0.73$; (6) burrow-dwelling flea infection rate: $a_6 = 0.97$; and (7) burrow-dwelling flea index: $a_7 = 0.76$.

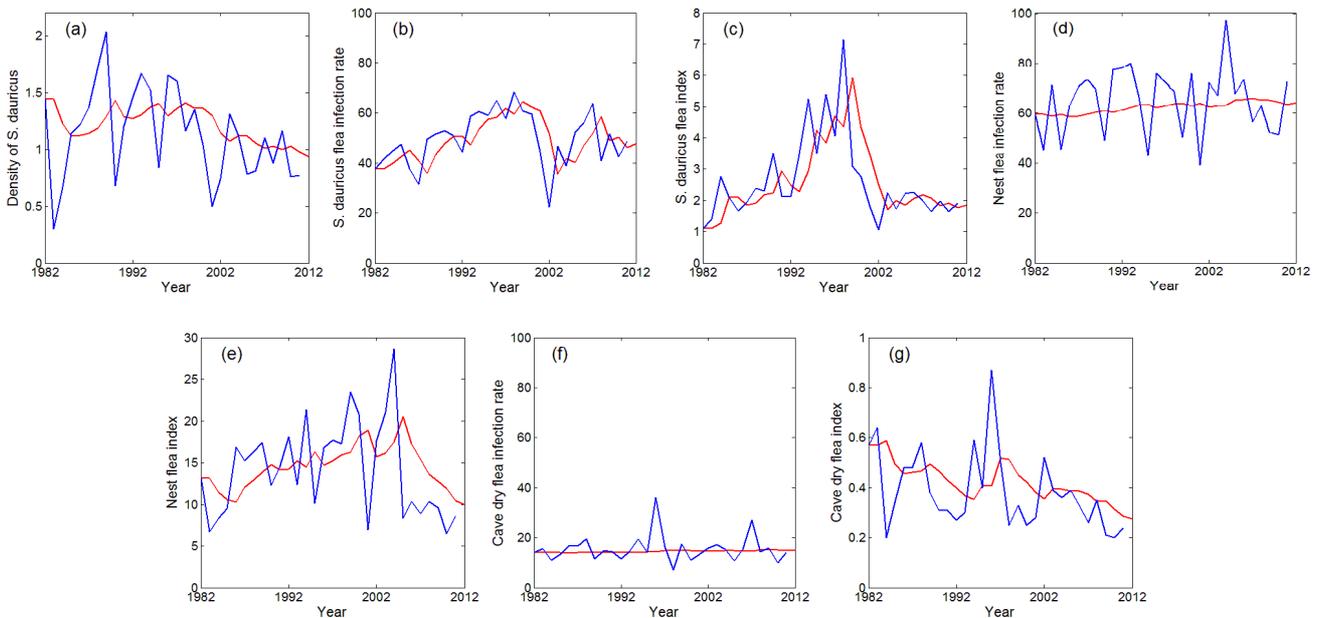


Figure 2. Exponential smoothing of the 7 main factors. Blue curves denote the observation values and red curves denote the exponential smoothing values. All the 7 factors show decreasing trend in recent years.

3. Results and Discussion

3.1. General Results

The exponential smoothing values of the 7 main factors in *S. dauricus* focus from 1982 to 2012 can be calculated by applying the exponential smoothing method introduced in Section 2.2.4. The trend graphs are shown in Figure 2. It is clear that in recent years, all the 7 factors show decreasing trend. This then gives us a first impression that the epidemic of *S. dauricus* plague is not likely to occur in 2012. The predicting values of the 7 main factors in 2012 are: (1) density of *S. dauricus*, $X_1(2012) = 0.94$; (2) *S. dauricus* flea infection

rate, $X_2(2012) = 48\%$; (3) *S. dauricus* flea index, $X_3(2012) = 1.85$; (4) nest flea infection rate, $X_4(2012) = 64\%$; (5) nest flea index, $X_5(2012) = 9.94$; (6) burrow-dwelling flea infection rate, $X_6(2012) = 14.88$; and (7) burrow-dwelling flea index, $X_7(2012) = 0.28$. From Table 1, y values of the best regression equations with numbers of factors 1-7 are $y_1(2012) = 0.183$, $y_2(2012) = 0.089$, $y_3(2012) = 0.085$, $y_4(2012) = 0.046$, $y_5(2012) = 0.050$, $y_6(2012) = 0.071$, and $y_7(2012) = 0.061$, respectively. According to the risk classification standard, all the y values are less than $1/3$. Therefore, our model predicts that no epidemic risk in 2012. This prediction is consistent with the actual situation: The plague bacillus *Yersinia pestis* was not detected in the *S. dauricus* focus in 2012.

3.2. Limitations of the Model

The combination of different statistical methods can help to get a more precise prediction, but it also requires more data in order to ensure all these methods are applicable. In general, establishing a reasonable regression model needs about ten observations for each state (two states in our model are “*Y. pestis* was detected” or “was not detected”) [32]. If the number of occurrences of plague in a region is not sufficient, then the results of regression method are questionable. Since *Y. pestis* was rarely detected in Jilin province, Liaoning province, and Heilongjiang province from 1982 to 2012, we cannot construct regression equations for these regions (although the risk of plague epidemic cannot be ignored). In contrast to regression method, the exponential smoothing method does not require any minimum number of observations. However, it can be used only for data without any systematic trend or seasonal components [28]. This implies that monthly data or quarterly data of the *S. dauricus* focus are not applicable for exponential smoothing method, we have to first transfer them to annual data in order to minimize the seasonal effects. On the other hand, the exponential smoothing value for a given year is derived from both the past smoothing values and the observation value of the year. Thus, if one observation value is missing, then it is unable to calculate the corresponding smoothing value (and also the subsequent smoothing values). A possible solution for this problem would be to use the Expectation-Maximization algorithm to estimate the missing value [33].

3.3. Potentials of the Model

It is worth to note that the key idea of our model could also be applied to predict the epidemic of plague in other foci. To achieve this, the first step is to find out the main influencing factors by regression methods; the second step is to establish a risk classification standard based on the history data and the regression results; the third step is to predict the trends of the main factors by time series methods; and the last step is to calculate the risk of epidemic in the next period. Depending on the data, one should choose suitable regression methods and time series methods. Previous studies have showed that different foci may have different main influencing factors, e.g., *M. unguiculatus* focus is sensitive to meteorological factors such as precipitation [20, 34]. Thus, epidemic predictions for these foci become even more delicate issues since there no universal time series method for meteorological data.

4. Conclusion

The regression methods (such as multiple regression) and the time series analysis (such as moving average methods) are commonly used for predicting infectious diseases. However, previous studies on plague prediction usually adopted a single statistical method to predict the occurrence of epidemic. In this paper, we establish a prediction model that combines the best subset regression method and the exponential smoothing method by the risk classification standard. Based on the data of the *S. dauricus* focus in the Inner Mongolia Autonomous

Region from 1981 to 2011, we constructed best regression equations with different numbers of influencing factors. We then derived the values of the 7 main factors in 2012 by using the exponential smoothing method, and calculated the risk of animal plague epidemic through the best regression equations. According to the risk classification standard, our model successfully predicted no animal plague epidemic in 2012.

Acknowledgments

This research received financial support from the Research Special Fund of Health Sector of China (No. 201202021), NSFC (No. 11301032) of China and “the Fundamental Research Funds for the Central Universities” of China.

References

- [1] Gage KL, Kosoy MY, 2005. Natural history of plague: Perspectives from more than a century of research. *Annu Rev Entomol* 50: 505-528.
- [2] Benedict CA, 1996. *Bubonic Plague in Nineteenth-Century China*. Stanford, CA: Stanford University Press.
- [3] Ben-Ari T, Neerinckx S, Agier L, Cazelles B, Xu L, et al., 2012. Identification of Chinese plague foci from long-term epidemiological data. *Proc Natl Acad Sci USA*, 2012, 109: 8196-8201.
- [4] Zhang G, Zhang GJ, Wang SR, Sun QT, Shan GG, 2002. Survey of natural focuses of plague and epidemic situations in China. *Chin J Ctrl Endem Dis*, 17: 101-103.
- [5] Hai R, 2011. The research advancement of plague natural foci in China. *Chin J Vector Biol & Control*, 22: 301-303.
- [6] Liu ZC, Zhang GJ, Zhang YB, 2000. Plague situation in China and prevention strategies [J]. *Chin J Ctrl Endem Dis*, 15: 281-283.
- [7] Li ZL, Baiyin MH, 1994. Mathematical models for forecast of population of *Marmota sibirica*. *Endem Dis Bull*, 9(4): 82-83.
- [8] Qing CY, Li ZL, 1995. Mathematical models for forecast of epizootic plague of *Citellus alaschanicus*. *J Ningxia Med Coll*, 17: 115-117.
- [9] Li ZL, Zhang WR, 1997. Ordos Mongolian gerbil plague prospective studies of animal disease forecast. *Chin J Endem*, 16: 202.
- [10] Li ZL, Zhang YX, 1998. Epidemic periodicity and dynamic forecast of plague of *Microtus brandti*. *Chin J Ctrl Endem Dis*, 13: 193-194.
- [11] Mi JC, Li ZL, Lv WD, Liu J, Wang LF, et al., 1997. Northern desert grassland of Inner Mongolia gerbil plague epidemic intensity forecast model. *Chin J Contl Endem Dis*, 12: 109-112.
- [12] Li ZL, Li SB, Zhou FX, 1998. Jilin province of *Spermophilus dauricus* population dynamics analysis. *Chin J Zool*, 33: 35-37.
- [13] Li ZL, Zhou FX, Li SB, Liu ZC, Liang BC, 2002. Mathematical models for forecast of epizootic plague of *Spermophilus dauricus* (VII). *Chin J Ctrl Endem Dis*, 17: 129-131.

- [14] Li ZL, Zhou FX, Li SB, Liu ZC, Liang BC, et al., 2003. Mathematical models for forecast of epizootic plague of *Spermophilus dauricus* (VIII). *Chin J Ctrl Endem Dis*, 18: 129-130.
- [15] Wang CG. 1998. Using fuzzy clustering analysis to predict *Spermophilus dauricus* epidemic animal disease dynamic. *Chi J of Endem*, 13: 193-194.
- [16] Ben Ari T, Neerinckx S, Gage KL, Kreppel K, Laudisoit A, et al., 2011. Plague and climate: Scales matter. *PLoS Pathog*. 7: e1002160. doi:10.1371/journal.ppat.1002160.
- [17] Zhou FX, Fang J, Liu ZC, 2003. Operation on use of information system in *Spermophilus dauricus* plague focus. *Chin J Ctrl Endem Dis*, 18: 100-103.
- [18] Xu C, Cong XB, Zhang GJ, Zhuang XW, Zhang YB, 2006. The Qinghai-Tibet railway plague epidemic situation and monitoring the development of geographic information system. *Chin J Ctrl Endem Dis*, 21: 153-154.
- [19] Lin XQ, Zhou SW, Tang CY, Liang JM, Deng QY, et al., 2011. Construction and application of geographic information system for plague in Guangxi. *Chin Trop Med*, 11: 1450-1451.
- [20] Zhang ZB, Li ZQ, Y Tao, Chen M, Wen XY, et al. (2007) Relationship between increase rate of human plague in China and global climate index as revealed by cross-spectral and cross-wavelet analyses. *Integr Zool*, 2: 144-153.
- [21] Xu L, Liu Q, Stige LC, Ben-Ari T, Fang X, et al., 2011. Nonlinear effect of climate on plague during the third pandemic in China. *Proc Natl Acad Sci USA*, 108: 10214-10219.
- [22] Xu L, Stige LC, Kausrud KL, Ben-Ari T, Wang SC, et al., 2014. Wet climate and transportation routes accelerate spread of human plague. *Proc R Soc B*, 281: 1471-2954.
- [23] Ji SL, 1990. Plague. Beijing: People's Medical Publishing House.
- [24] Gong SL, Liu ZC, 2007. Research Status of Plague Natural Foci of Jilin Province. Jilin: Jilin People's Publishing House.
- [25] Hocking RR, Leslie RN, 1967. Selection of the best subset in regression analysis. *Technometrics*, 9: 531-540.
- [26] Li ZL, Wang CG, 1992. Two mathematical models for forecast of plague epizootic of *Citelas dauricus* (V). *Endem Dis Bull*, 7(3): 50-52.
- [27] Zhou XL, Zhang BY, Cong XB, Li ZL, Yao XH et al., 2013. Early-warning index system of animal plague in *Spermophilus dauricus* focus. *Chin J Ctrl Endem Dis* 28: 401-403.
- [28] Brown, Robert G. 1956. Exponential smoothing for predicting demand. Cambridge, Massachusetts: Arthur D. Little Inc. p.15.
- [29] Zeng ZY, Yang YM, Song ZM, 1994. A comparative study of biomass dynamics of the species populations of the desert community in the Chihuahuan desert of North America. *Zool Res*, 15: 32-41.
- [30] Li SB, Ma YK, Li C, Fan MG, Zhang YB, 1999. Animal Plague Epidemiology. Jilin: Editorial Department of Chinese Journal of Endemic Disease Prevention and Control.
- [31] Zhou XL, Zhang BY, Cong XB, Li ZL, Yao XH et al., 2014. Application of best subsets regression on the risk classification for *Spermophilus Dauricus* focus. *Chin J Epidemiol*, 35: 62-65.
- [32] Harrell FE 2001. Regression Modeling Strategies: with Applications to Linear Models, Logistic Regression, and Survival Analysis. Berlin: Springer-Verlag.
- [33] Shumway RH, Stoffer DS, 1982. An approach to time series smoothing and forecasting using the EM algorithm. *J Time Anal*, 3: 253-264.
- [34] Li ZL, Li SB, Zhang WR, Zhang YX, Zhou FX et al., 1999. Studies on early warning systems of three species animal plagues in northern part of China. *Chin J Ctrl Endem Dis*, 14: 193-196.