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**Modelling of PM₁₀ concentration for industrialized area in
Malaysia: A case study in Shah Alam**

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Abstract

In Malaysia, the predominant air pollutants are suspended particulate matter (SPM) and nitrogen dioxide (NO₂). This research is on PM₁₀ as they may trigger harm to human health as well as environment. Six distributions, namely Weibull, log-normal, gamma, Rayleigh, Gumbel and Frechet were chosen to model the PM₁₀ observations at the chosen industrial area i.e. Shah Alam. One-year period hourly average data for 2006 and 2007 were used for this research. For parameters estimation, method of maximum likelihood estimation (*MLE*) was selected. Four performance indicators that are mean absolute error (*MAE*), root mean squared error (*RMSE*), coefficient of determination (*R*²) and prediction accuracy (*PA*), were applied to determine the goodness-of-fit criteria of the distributions. The best distribution that fits with the PM₁₀ observations in Shah Alam was found to be log-normal distribution. The probabilities of the exceedences concentration were calculated and the return period for the coming year was predicted from the cumulative density function (*cdf*) obtained from the best-fit distributions. For the 2006 data, Shah Alam was predicted to exceed 150 µg/m³ for 5.9 days in 2007 with a return period of one occurrence per 62 days. For 2007, the studied area does not exceed the MAAQG of 150 µg/m³.

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• **Introduction**

Exponential development of science and technology nowadays has lead to the rapid growing industrialization which is the major sources of various environmental pollutions, especially air pollution. Airpollutants, specifically particulate matter (PM) smaller than about 10 micrometers, referred as PM₁₀, have received extensive attention, due to its capability to settle in the bronchi and [lungs](#) and cause health problems [1]. Malaysian Ambient Air Quality Guidelines (MAAQG) were issued and target values for annual and daily mean mass concentrations for various air pollutant were established to control and reduce air pollutant levels in the atmosphere. Monitoring data and studies on ambient air quality show that some of the air pollutants in several large cities are increasing with time and are not always at acceptable levels according to the MAAQG. There are very limited data and case studies on air pollution in our country. Most of the air modeling using probability distribution is only applied in foreign countries.

Statistics are important in the analysis and interpretation of data in which the outcomes from the analysis can be utilized as prediction tools that have become the major aim in environmental engineering [2]. There are many statistical procedures to analyze various environmental data sets which are frequently asymmetrical and skewed to the right (that is with long tail towards high concentrations) [3]. Many types of probability distributions have been used to fit air pollutant concentrations including Weibull distribution [4], lognormal distribution [5], gamma distribution [6], Rayleigh distribution [7], Gumbel distribution [8] and Frechet distribution. Lu [9] and Chen *et al.* [10] have studied the goodness-of-fit for selected probability distributions by using several performance indicators such as mean absolute error (*MAE*), root means error (*RMSE*), index of agreement (d_2), bias (*B*), normalized absolute error (*NAE*), prediction accuracy (*PA*) and coefficient of determination (R^2). The goals of this research were to study the statistical characteristics of the observed data, as well as to select the best-fit distribution in order to predict the exceedences and return period of the PM_{10} critical concentration.

• Data and methods

2.1. Data set

- The datasets consisted of PM_{10} concentration on a time-scale of one per hour (hourly averaged) for 2006 and 2007 in Shah Alam, Selangor. It is an industrialized area with high population and traffic density with the weak prevailing winds was recorded causing the air contaminants to stagnate [11].

2.2. Probability distributions

- Six theoretical distributions, namely Weibull, gamma, log-normal, Rayleigh, Gumbel and Frechet distributions are used to fit the entire measured PM_{10} data [12,13]. For parameters estimation, method of maximum likelihood estimation (*MLE*) was selected.

2.3 Performance indicators

- Four performance indicators (PI) that are mean absolute error (*MAE*), root mean squared error (*RMSE*), coefficient of determination (R^2) and prediction accuracy (*PA*), were applied to determine the goodness-of-fit criteria as to judge which type of parent distribution is the most appropriate to represent the PM_{10} pollutant concentration [14].

2.4 Exceedences and return period

- Once the best-fit distribution is determined, the cumulative distribution function (cdf) of the fitted distribution was used to calculate the exceedence, or the probability that the event is equalled or exceeded in computed period. The reciprocal of the exceedence probability was calculated so to obtain the return period (also known as the recurrence interval) of the event.

• Results and discussion

• 3.1 Data description

• Table 3.1 gives the summary of the descriptive statistics for PM₁₀ hourly data of Shah Alam for 2006 and 2007. The mean values for the area in both years are higher than their respective median which indicates that the pollutants distributions are positively skewed (also called right-skewed). This means most of the data is concentrated on the left of the figure with few high values. The maximum value for Shah Alam for 2006 was 313.0 decreased to below MAAQG limit of 150 µg/m³ in 2007. Fig. 1. indicates the PM₁₀ concentrations had small exceedences starting from the end of September to the mid of October due to the haze episodes in Malaysia in 2006 [15]. No exceedence above 150 µg/m³ was observed for 2007 as shown in Figure 1.

3.2 Probability distributions

Table 2 shows the parameter estimates of the six distributions for 2006 and 2007. All the estimates have been obtained using maximum likelihood estimators (MLE). It is observed that the value of scale parameter, β, is always larger than the corresponding shape parameter, α, indicating the computations are done right.

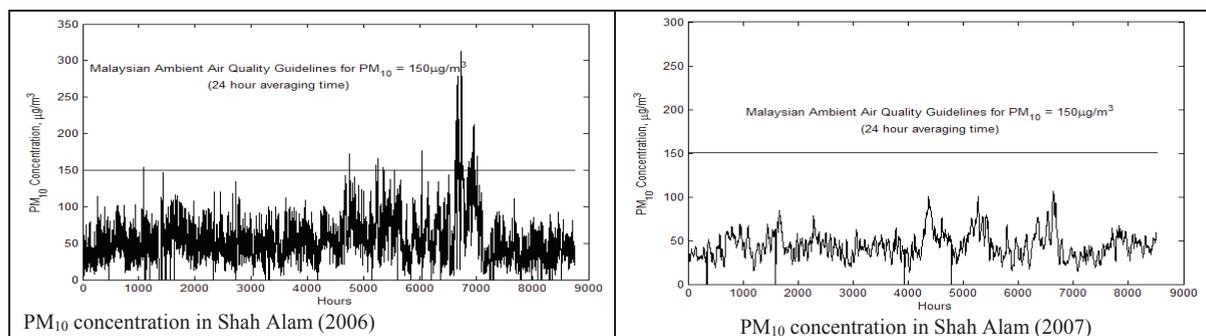


Fig.1. Time series plot of PM₁₀ concentration in Shah Alam for 2006 and 2007

Table 1. Descriptive statistics for PM₁₀ concentration

	Shah Alam		
	2006	2007	
Valid Data	8598	8462	
Missing Data	162	298	
Mean	55.7	44.5	
Median	50.0	42.8	
Standard Deviation	30.7	14.6	
Mode	48.0	41.7	
Variance	942.8	212.6	
Skewness (Standard Error)	2.13 (0.03)	0.86 (0.03)	
Kurtosis (Standard Error)	8.79 (0.05)	1.23 (0.05)	
Minimum Value	6.0	13.2	
Maximum Value	313.0	106.9	
Range	307.0	93.7	
Percentiles	25	36.0	34.2
	50	50.0	42.8
	75	69.0	52.4

Table 2. Parameter estimates

Distributions	2006		2007	
	α	β	α	β

Weibull	1.933	63.06	3.167	49.622
Gamma	3.921	14.216	4.603	9.67
Log-Normal	0.524	3.888	0.327	3.743
Rayleigh	-	44.999	-	33.118
Gumbel	21.34	42.959	12.0	37.774
Frechet	1.739	37.3	2.936	35.756

Fig.2 shows the cumulative distribution (cdf) plots for PM₁₀ concentration in Shah Alam where six distributions were plotted and compared with the observed distribution.cdf plots from For 2006, gamma, log-normal and Gumbel distributions have better fit on the PM₁₀ observation data in Shah Alam. For 2007, log-normal distribution has better fit than Weibull, gamma, and Gumbel which also fit well the observed data. Frechet distribution overestimates at concentration less than 27 µg/m³ and underestimates after that indicating the worst fitting.

• 3.3 Performance indicators

- Table 3 shows the smallest value for *MAE* is given by the Gumbel distribution (2006) and that of *RMSE* is given by log-normal distribution which also indicates the highest values for *R*² and *PA*, i.e. 0.989 and 0.995 respectively. Gamma distribution (2007) gives the smaller value of *MAE* that is 0.5521 compared to log-normal distribution which is 0.6631. However, log-normal distribution indicates smaller value of *RMSE* that is 1.0561 compared to gamma distribution, 1.0938. The highest value for *R*² and *PA* are both given by log-normal distribution with the value of 0.9953 and 0.9978 respectively. Based on the analysis of these results, log-normal distribution fits the data better than Gumbel distribution (2006) and gamma distribution (2007) in representing the PM₁₀ concentration in Shah Alam for both 2006 and 2007.

• 3.4 Exceedences and return period

The distribution that fits the PM₁₀ concentration in Shah Alam is log-normal for both 2006 and 2007. From Fig.4, the probability that the PM₁₀ concentration for 2006 equal or less than 150 µg/m³ is 0.9839 [that is, $P\{X \leq 150\} = 0.9839$] and the probability that the concentration greater than 150 µg/m³ is 0.0161 [that is, $P\{X > 150\} = 0.0161$]. There will be 5.9 days where the PM₁₀ concentrations in 2007 exceed 150 µg/m³. Hence the return period for 2007 is once per 62 days. Meanwhile, the probability that the concentration in Shah Alam for 2007 greater than 150 µg/m³ is 0 [that is, $P\{X > 150\} = 0$]. This shows the PM₁₀ concentrations for the whole year stay below 150 µg/m³. There is no return period predicted for concentration above 150 µg/m³ in 2008.

• Conclusion

- Six distributions namely Weibull, log-normal, gamma, Rayleigh, Gumbel and Frechet distributions were chosen to model the PM₁₀ observations in Shah Alam, Selangor. One-year period hourly average data for 2006 and 2007 were used. Method of maximum likelihood estimation (*MLE*) was selected for parameters estimation. Four performance

indicators specifically mean absolute error (*MAE*), root mean squared error (*RMSE*), coefficient of determination (R^2) and prediction accuracy (*PA*), were applied to determine the goodness-of-fit criteria of the distributions. The best distribution that fits with the PM_{10} observations in Shah Alam was found to be log normal distribution. The probabilities of the exceedences concentration were calculated and the return period for the coming

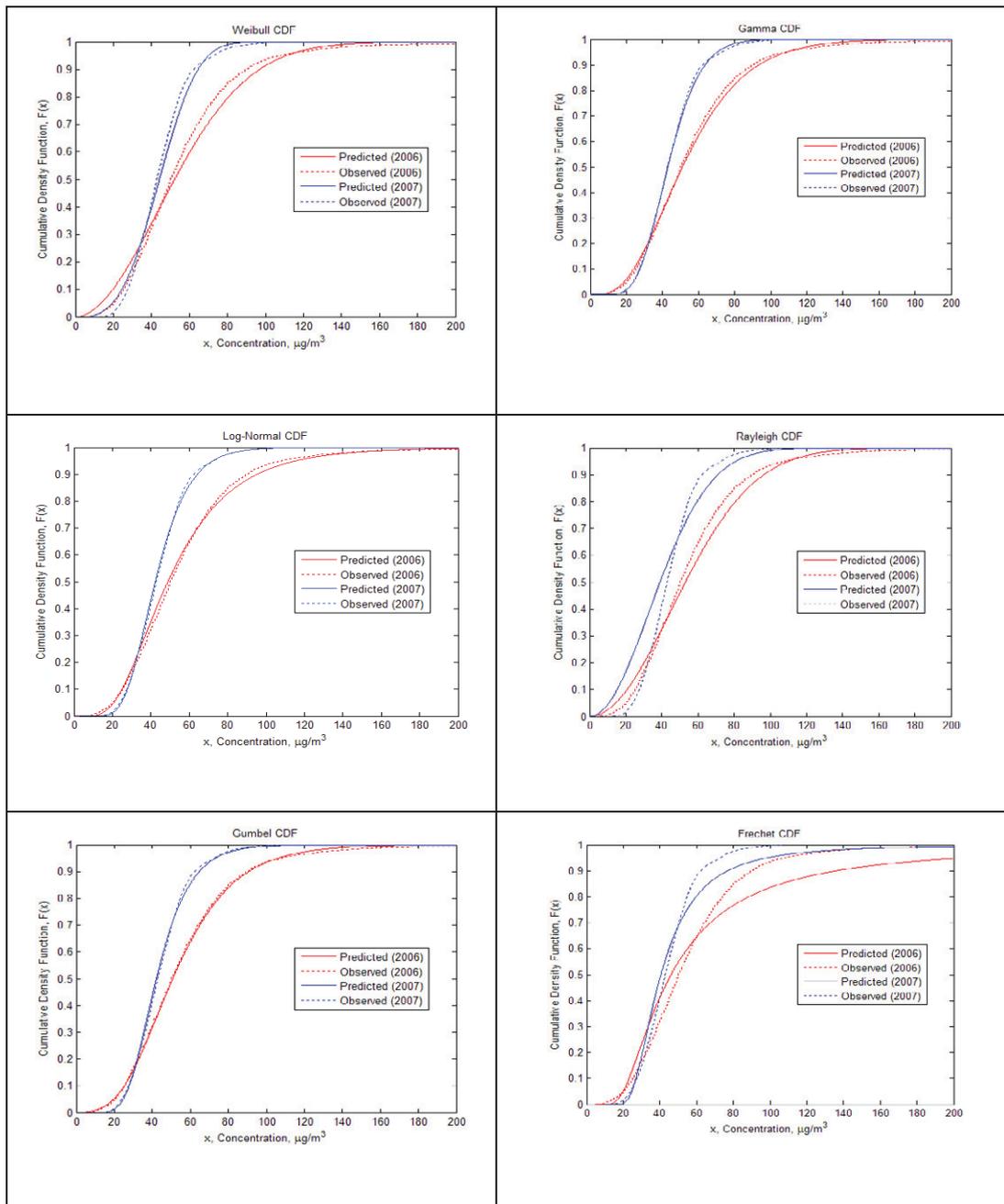


Fig.2. cdf plots for PM_{10} in Shah Alam

Table 3. Performance Indicators value for PM_{10} concentration in Shah Alam

Distributions	Performance Indicators			
	Mean Absolute Error (<i>MAE</i>)	Root Mean Squared Error (<i>RMSE</i>)	Coefficient of Determination (R^2)	Prediction Accuracy (<i>PA</i>)

	2006	2007	2006	2007	2006	2007	2006	2007
Weibull	4.758	2.174	8.296	2.839	0.927	0.967	0.963	0.983
Gamma	2.253	0.552	6.555	1.094	0.958	0.994	0.979	0.997
Log-Normal	2.009	0.663	3.267	1.056	0.989	0.995	0.995	0.998
Rayleigh	4.686	6.829	8.541	7.88	0.923	0.991	0.961	0.996
Gumbel	1.514	0.823	6.111	1.39	0.969	0.994	0.985	0.997
Frechet	25.063	6.641	133.26	21.241	0.528	0.74	0.727	0.86

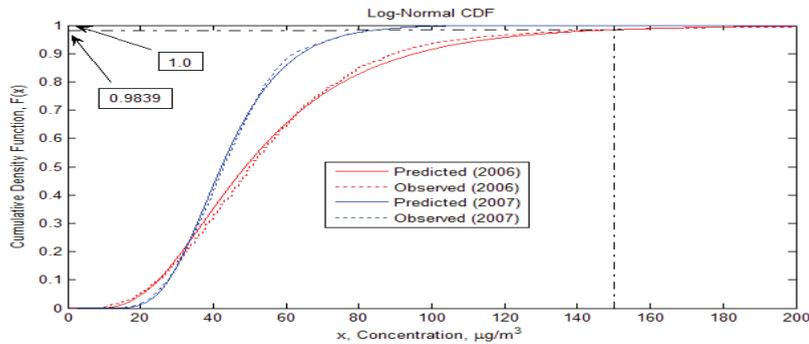


Fig.3. Estimation of exceedences above MAAQG (150 µg/m³) in Shah Alam for 2006 and 2007 using log-normal cdf plot

year was predicted. For the 2006 data, Shah Alam was predicted to exceed 150 µg/m³ for 6 days in 2007 with a return period of one occurrence per 62 days. However, the PM₁₀ observations for 2007 do not exceed the MAAQG of 150 µg/m³.

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