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Research paper

Socio-Environmental Factors and Tuberculosis: an Exploratory Spatial Analysis in Peninsular Malaysia

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Abstract

Spatial pattern of tuberculosis (TB) describes the environmental variation of the disease. Controlling the disease is not only depends on bio-medical method, but there is also a need to explicitly understand socio-environmental contexts which include the elements of people, space and time. A spatial exploration of local TB dynamic in Malaysia's perspectives is necessary since earlier studies were limited in addressing these contexts. This paper discusses a general spatial pattern of TB distribution pattern in Peninsular Malaysia for a ten-year period from 2005 to 2014 and its general possible correlation with socio-environmental factors in 2010. Geographical information system (GIS) and correlation analysis are fundamental techniques used to explore the local pattern of TB distribution and the relationship. Data on annual TB cases and state map were collected from the Ministry of Health and the Department of Rural and Country Planning Malaysia respectively. The overall spatial pattern of TB cases has shown increasing trends and concentrating in five states, including Selangor, Johor, WPKL, Kelantan and Kedah, even though there is no specific pattern in the area. This study suggests that the majority of the cases have occurred in the urban states, having high-medium incomes and populous areas.

Keywords: *GIS Correlation; Peninsular Malaysia; Social Environment.; Spatial Pattern;Tuberculosis*

1. Introduction

This TB occurs in every part of the world and can kill three million people in the world for both of developed or developing countries. [8] reported the majority of the TB cases come from the 22 high burden countries, especially from African and Asian regions. Malaysia recorded medium burden cases of TB incidence with 81 per 10,000, 24,711 cases, and 1,603 deaths in 2004, but the number of new TB cases increased from 15,000 in 2005 to 19,251 in 2011 **(**MOH, 2012). These local situations and dynamics bring up key research question: which states have the highest cases of TB? and why are the states experienced such risk situations?

Previous studies have shown the environmental risk factors in high burden countries of TB can be divided into two main indicators, including ecological environment, human or social environment. The ecological environment is another term of the environment, relating the biotic and abiotic surrounding of an organism or population, For example, TB high-risk prevalence areas are co-impacted by spatial proximity or geographical factor [1][2][12] especially in urban area [1][12][3][38], poor housing quality, crowded and small living condition [14]. Low SES also contributes to the global TB occurrences [2][12] such as unemployment, low educational level and poverty. Human factors both individual and population have been also identified as a major of global TB epidemics, especially in crowded or higher population, migrant population, high risk group or people with co-infection or infectious disease such as HIV, drugs abuse, AIDS and so on.

Therefore, molecular and analytical epidemiology used to control the disease in terms of local demographic characteristics, genetic, behaviours, environmental exposures, and other potential risk factors. However, there is a need of a geographically based tool for a better understanding of the TB epidemiology (Murray and Alland, 2002; Narayanan, 2004). Studying spatial pattern of TB in Malaysia is imperative to understand well the spatial distribution of the disease since there are limited studies on these particular local perspectives A spatial analysis framework developed by [37] is general guidelines used to control disease in medical and geographical perspective by identifying a spatial pattern and of disease cluster, as well as to predict disease risk factors and areas. This analysis can stimulate an idea to create a hypothetical framework for the dynamics TB distribution pattern and risk factors complexities in the context of Malaysia

2. Review on Socio-Spatial Relations in the Tuberculosis Environment

2.1. Geographical Pattern of Tuberculosis Distribution

Tuberculosis (TB) is caused by tuberculosis bacilli called as mycobacterium tuberculosis that is spread from person to person through the air, and then it can transmit through a community across the geographical boundaries. TB Epidemiology is a study of the disease distribution and determinants of risk factor in a specified population, and to the control of health problems that related with time, place, and person characteristics. Narayanan, (2004) has suggested apart of human-based tool, there is a need of

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the geographically based tool for a better understanding and controlling of tuberculosis cases.

TB remains a major global health problem that can attack any people or regions both in developed or developing countries. [7] described global TB incidence and estimates that approximately one-third of the global community is infected with M. tuberculosis. TB is the second most common cause of death due to an infectious disease after human immunodeficiency virus HIV/AIDS. In 2013, the majority of the TB cases come from African and Asian regions especially in South Africa and India [8], where the largest number of new TB cases occurred in the South-East Asia and Western Pacific Regions, accounting for 56% of new cases globally.

In Malaysia, the TB among high risk groups continue to be an important disease [18]. The number of new TB cases in the country increased from 15,000 in 2005 to 19,251 in 2011 especially in three states of Sabah, Selangor and Sarawak. While pulmonary TB is the commonest form of TB in Malaysia, extrapulmonary TB (EPTB) still posed a threat. The majority of patients are in the 21 - 60 years age group (69.5%) and there is a male predominance (65%). Overall, most of the global TB cases occur in Asian and African regions, especially in developing countries. In Malaysia, the clustering is dynamically cases since every state has owned risk factors and potentials to be attacked by the TB cases.

2.2. Tuberculosis in Socio-Spatial Dimensions

Empirical global studies have shown the common environmental (endogenous or extrinsic) risk factor affects TB risk vulnerability are the biophysical or ecological environment, socioeconomic status, human population and demographic characteristic. TB spatial clustering and variation, the high-risk areas are co-impacted by areal proximity or geographical factor [1][2][12] especially in urban area [1][2][3], and socio-economic status [2][3].

Moreover, human factor both population or individual have been also identified as a major of global TB epidemics including crowded or higher population [4][38], migrant population, high risk group and people with co-infection or another infectious disease such as HIV, drugs abuse, AIDS and so on. Since each country has unique local environments, then the effect of risk factors and dynamics of TB occurrences are might be different in particular areas. As a result, some experts have pointed out some recommendations on these particular problems, as well as to enhance the current result.

The well-established risk factors need to combine among themselves or other potential intrinsic risk factors for major impact on the state of public health in the current century [10] and better reflect the true risk factors of TB burden and guide national TB programmes to execute more effective TB control interventions. Moreover, there a need to be studied in depth in future research for complex relationships such as between SES and environment [1][13][15][33][11][12][16][10][17] .

In the context of Malaysia, The [12] identified the risk of TB transmission came from a high risk group. This human or demographic factor may the main factor affecting local TB occurrences such as pulmonary TB (PTB), patients are in the 21 - 60 years age group and male predominanc [12]. Table 1 summarises the common risk factors according to global situations, focusing on social economic status, human population and demographic. It can be concluded that in order to determine potential key risk factors of TB in certain local areas, studying spatial pattern and precise causal associations between TB cases and related risk conditions is crucial to investigate the real scenario of local TB cases.

Table 1: Common Risk Factors in Global Tuberculosis

Risk Factor	Specific criteria/determinants	Author (s), region (s)
1. Socio Economic Status, SES (Job, salary and occupational-related to poor and social vulnerability)	**1.1** The level of education of people with primary level or with illiterate level, type of TB and place of residence **1.2.** Habitat/settlement censuses, and living conditions **1.3.** TB is linked with SES environmental factors such as living condition and ecological condition especially in urban area **1.4.** Linked to both the quality of health care and socio-economic status. Confirmation of TB cluster areas may help public health authorities to set up priorities regarding to be targeted for prevention or control measures **1.5**. Effects of social vulnerability on the disease, complex relationships may exist between TB incidence and a wide range of environmental and intrinsic factors, which need to be studied in depth in future research **1.6**. TB situations are related with in neighbourhoods of low SES	**1.1** [19], West Azerbaijan Province, Iran. **1.2.** [19][20][22][21] **1.3.** [1][12][13][10][13][33][15][16] **1.4.** [2] Antananarivo, Madagascar. **1.5** [11], Brazil**.** **1.6.**[23][24] Cameroon and Southern Africa.
2 Population and Demographic Characteristics (crowded/density environment, migration and low SES	**2.1** Geographical characteristics of the area at the risk are high density/ populated areas in the urban and rural areas, poor sanitation, poorly built housing, and lack of both adequate sewage systems and water supply facilitate pathogen spread. These situations are related with in neighbourhoods of low SES and HIV/TB mortality among children aged 1-5 years **2.2** TB related with socio-economic status specially poverty, urbanicity and population density in comparison with rural areas, but the higher level of CS/location, risk factor/biological and comparative studies need to be considered.	**2.1** [12][10][13][33][15][16], African Region**.** **2.2** [27[28[29][30][31]American Latin. **2.3** [32][33][34][35] **2.4** [33], USA.

	2.3 The high proportion of recent TB transmission also strong related with the mobility of population especially from other country. Mobility people from other countries to other will bring the transmission regionally and local people as occurred in USA 2.4 The characteristics of high risk area are the countries of birth of foreign-born persons especially from Mexico and Philippines, India and Vietnam: • Race and Ethnicity- Asians had the highest TB rates • HIV-Infected Persons- HIV-infected persons are at high risk for developing • Multidrug-Resistant TB-Identifying Groups at High Risk for TB	

3. Methodology

The analytical framework of GIS and spatial epidemiology adapted from [37] is the research methodology used in this study. It includes two main stages, namely, data collection, data processing and data analysis. The first stage describes the study area and datasets used in the study while the second stage covers the process and analysis of the data input of cases and selected variables using classification and Correlation and Spatial Analyst tool in ArcGIS. The data analysis describes spatial pattern of TB cases in the Peninsular Malaysia from 2005 to 2014.

3.1. Study Area and Data Collection

Peninsular Malaysia or West Malaysia, in Malaysia, is selected as the study area. Malaysia is located in Southeast Asia, with Kuala Lumpur as its capital city (Fig. 1). The Peninsular Malaysia has an area of 8.10 million ha, consisting of twelve administrative states from the smallest area (Perlis), to the largest area (Pahang). The TB cases of state of Wilayah Persekutuan Putrajaya (WPP) are combined into Wilayah Persekutuan Kuala Lumpur (WPKL) because of geopolitical changes. The area as a whole is composed of 80% present lowland and 20% highland.

Annual cumulative TB cases from 2005 to 2014 and outline state map of the study area were collected from the TB/Leprosy Sector in Ministry of Health (MOH) and the Department of Town and Country Planning (JBPD), Malaysia. General attributes of the population and urban population (people), monthly income (RM) and non-forested land (km^2) in 2010 data were acquired from available online sources from the Department of Statistics, Economic Planning Unit (EPU) and Forestry Department in Malaysia. These secondary non-spatial data then were spatially converted and processed into ArcMap and ArcCatalog using a quantitative classification and a correlation coefficient technique.



Fig. 1: Map of Peninsular Malaysia, Malaysia.

3.2. Data Processing, Calculation and Analysis

The selected data were processed using spatial software or ArcGIS developed by the Environmental System Research Institute (ESRI), especially ArcMap and ArtToolbox tools. The software has some important GIS functions for geocoding, thematic mapping or displaying the data using Malayan Rectified Skew Orthomorphic (MRSO) coordinate systems. Basic GIS operations such as file setup, joining excels spreadsheets with geographic ID fields, and thematically representing data on the maps were conducted in ArcMap and Spatial Analyst tool.

ArcMap is the primary application in ArcGIS and to perform a wide range of common GIS tasks. The mean value of the cumulative TB cases, and the selected variables were used as a scale for determining the risk level of TB occurrences and risk factors in this study. For example, the total TB cases in the study area are 136, 864 people and its mean value is 11, 405 people after they are divided by 12 (states). If the total of cases in a particular state is more than the mean value, hence the state is assumed as a high risk area (darker colour).

Pearson's correlation coefficient, *r* is also another statistical technique applied in this study to measure the descriptive pattern, association and strength of the relationship between any two variables such as environmental factors and TB cases (Fig. 2). [36] states if the relationship is weak, then knowing the value of one attribute variable does not predict the value of the second variable.

The value of this correlation is also expressed in terms of a coefficient and it has a range (maximum negative [-1] and positive [+1]). When the correlation coefficient approaches a maximum negative value, it means that there is a strong negative correlation. This means that the higher the value of one attribute variable, the lower the value will be of the other variable (e.g. the lower the monthly income rate, the higher the TB incidence). There are two ways to calculate the correlation, which are based on sample data (Fig. 2a) and means and standard deviations (Fig. 2b).

(a)

$$r = \frac{n\sum_{i=1}^{n} x_i y_i - \left(\sum_{i=1}^{n} x_i\right)\left(\sum_{i=1}^{n} y_i\right)}{\sqrt{\left[n\sum_{i=1}^{n} x_i^2 - \left(\sum_{i=1}^{n} x_i\right)^2\right]\left[n\sum_{i=1}^{n} y_i^2 - \left(\sum_{i=1}^{n} y_i\right)^2\right]}}$$

(b)

$$r = \frac{\frac{\sum_{i=1}^{n} x_i y_i}{n} - (\bar{x})(\bar{y})}{s_x s_y}$$

n = sample size of x and y data $\bar{x}\ \bar{y}$ = means of x and y data

$s_x s_y$ = standard deviations of x and y data

Fig. 2: Calculation of Pearson's correlation coefficient, *r* [36]

4. Result and Discussion

4.1. Spatial Pattern of Tuberculosis Cases

Fig. 3 and Fig. 4 show the total cumulative TB cases by states in Peninsular Malaysia for 10 years period from 2005 to 2014. The total number of the cumulative cases recorded for the 10 years were 136, 864 cases, in which Selangor dominated the cases with 21.7 % of the grand total of the cases, followed by Johor (13.7%), WPKL (11.3%), Perak (9.8%), and Kelantan (9.7%). While, Pulau Pinang, Kedah, Pahang and Terengganu recorded the medium risk of TB cases, started from 7.6% to 5.1%. The other states that reported the total number of TB cases less than 5,000 cases or 5% were considered as low risk states, consisting of Negeri Sembilan (3.5%), Melaka (3.2%), and Perlis (0.9%).

The cases which notified more than the mean value of the total cumulative TB cases or 11,405 cases (8.3%) are categorised as an abnormal situation or a high risk state, whilst the others are identified as medium and low risk states. Therefore, the figures also reveal that more than 66.2% of the overall cases in the area were assumed as high risk states, including Kelantan, Perak, WPKL, Johor, and Selangor. The other states only expressed low and medium risk situations, especially in Pulau Pinang, Kedah, Pahang Terengganu, Negeri Sembilan, Melaka and Perlis.

Referring to the TB time series plot by the states and years in Peninsular Malaysia from 2005 to 2014 (Fig. 5), it is clearly explained that Selangor dominated the total cases by 21.7 %. The other states reported constant higher cases than the mean value (1141 cases) were Johor, WPKL, Perak and Kelantan. The rest of the states only showed the medium and low risk of TB cases, particularly in Perlis, Melaka and Negeri Sembilan.

An interesting trend was found in Selangor, Johor, WPKL, Pulau Pinang and Kedah where they have more significant dynamics or variations of cumulative TB distribution compared to the other states. These phenomenal dynamics might be caused by several possible factors, including the local complexities of TB transmission process, or the effectiveness of the particular state health TB control program in the disease detection and treatment.

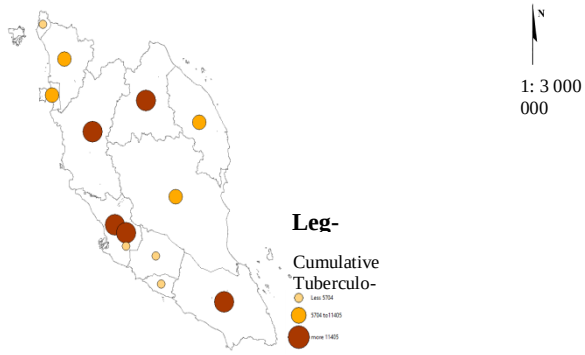


Fig. 3: A spatial pattern of cumulative TB cases in Peninsular

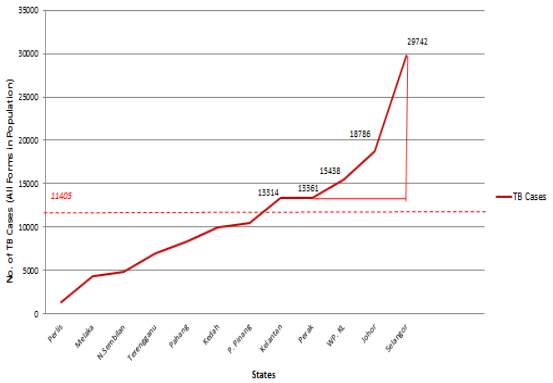


Fig. 4: A Trend of cumulative TB cases in Peninsular Malaysia, 2005-2014

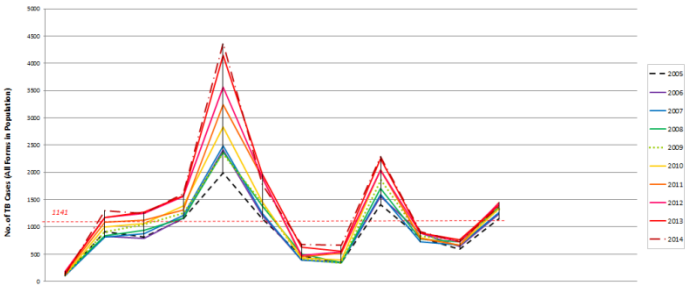


Fig. 5: Trends of TB cases in Peninsular Malaysia from 2005 to 2014

Fig. 6 illustrates the spatial distribution pattern of TB cases in Peninsular Malaysia by states using geographical information system (GIS) technology. Similarly, the cases increased in term of a number of cases and every state had also experienced with TB occurrences. In the early three years of 2005 to 2011, the cases are likely to be significantly increased in some regions of the study area, especially in Selangor, WPKL, Johor, Perak and Kelantan. After that, in 2012 to 2014, the cases spread out to Kedah and P.Pinang then consistently occurred in these states. The state of Perlis, N.Sembilan and Melaka only notified low-medium risk for ten year-cases

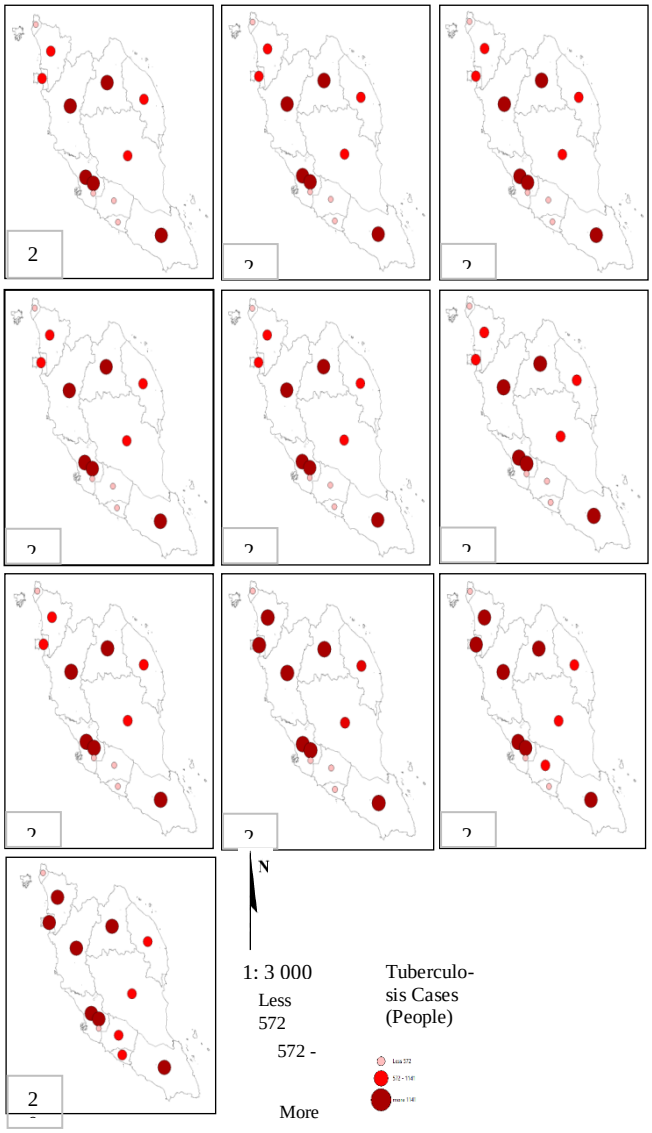


Fig. 6: A spatial distribution pattern of TB cases and risk level in Peninsular Malaysia

There was no specific clustering found in the study area since the pattern was slightly randomly distributed. The states with high risk situations have similarities in term of the environmental characteristics of TB ecology, particularly in the case of population changes. However, there were some differences in terms of biophysical environments and socio-economic status (SES) since TB could be occurring either in urban and/or in rural areas even though they are having high urbanisation or better socio-economic status. For example, Kelantan and Perak are in medium category in term of the socioeconomic status and urbanization, compared to Selangor and WPKL. This result is consistent with finding found from previous studies that geographical characteristics of the area at the risk are high density/ populated areas in the urban and rural areas [10][23][24][25][26].

Table 2: Potential risk states in Peninsular Malaysia

Risk Level of TB (Cases)	States	Year
Cases High	Selangor, Johor, WPKL, Perak Kelantan	2005-2014
(All Forms) Medium	P. Pinang, Kedah, Pahang, Terengganu	2005-2014
Low	N.Sembilan, Melaka, Perlis,	2005-2014

Table 3: Correlation analysis of TB risk factors in Peninsular Malaysia

TB Clustering	Correlation	Range of Period
Overall TB Clustering of Cases	No apparent clustering is detected at this scale (Random)	2005-2014
Correlation (2010)		
Possible Effect of Related Environments on TB Cases		
TB Cases Vs No. of Population	r = 0.953 (Very Strong)	2010
TB Cases Vs Mean Monthly Household Income	r = 0.578 (Moderate)	2010
TB Cases Vs Urban Population	r = 0.403 (Weak)	2010
TB Cases Vs Non Forested Land	r = -0.134 (Very Weak)	2010

4.2. Spatial Correlation between Tuberculosis Cases and Socio-Environmental Factors

The results of the trendlines of TB in Fig. 7 revealed the cases in Peninsular Malaysia seemed to potentially occur at anywhere and any local environments, especially in urban and populous areas of human socioeconomic or recreational concentration. For example, Selangor, Johor and WPKL are located in urban areas and have high income states, where their Mean Monthly Gross Household Income (EPU, 2015) are more than the mean value of the overall monthly income of the study area in 2010, showing human factor has a significant effect on TB cases in the states. Statistically, correlation analysis also found that TB cases in 2010 were significantly correlated with the number of population and mean monthly income (r > 0.500) rather than urban population and non-forested land (r <0.500) (Figure 22). This correlation exposed the impact of local risk of of the population and social economic status could be higher than biophysical environments on the TB occurrences.

4.3. Spatial characterisation and correlation of potential risk tb states in Peninsular Malaysia

Table 2 and Table 3 highlight the key findings in this study, comprising three main level of TB risk describing the TB spatial pattern from 2005 to 2014. Obviously, result from the GIS and spatial epidemiological techniques have demonstrated the spatial distribution pattern of TB cases in Peninsular Malaysia was randomly distributed since there were no clear qualitative clustering have been detected. But the state of Selangor, Johor and WPKL were still the top states in recording the highest cases. These dynamics of TB occurrences can lead to the local complexities in determining explicit risk factors, causing the cases could occur at any place and biophysical environments in Peninsular Malaysia especially in Selangor, WPKL, Johor, Melaka and Pulau Pinang.

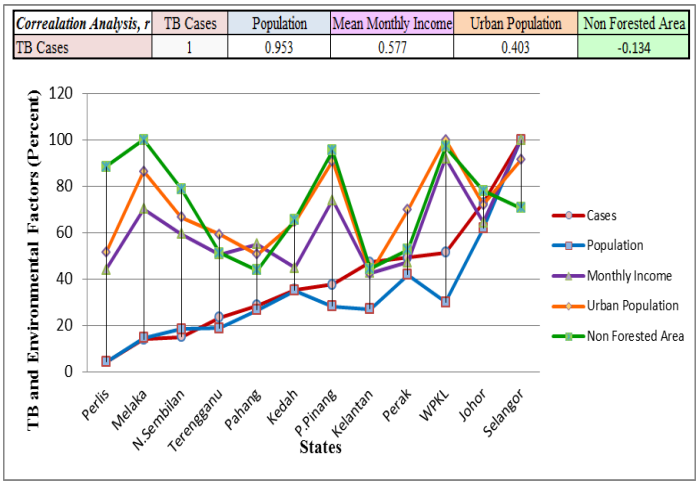


Fig. 7: Correlations between TB cases and selected risk factors in Peninsu-

5. Conclusion

An explicit understanding on the TB distribution pattern is a fundamental process to explore an initial idea of the local transmission of diseases, the possible relationship with related environmental risk factors, and identifying potentially high risk areas. The integration of geographical information system (GIS) and spatial visualisation techniques have demonstrated their analytical capabilities for exploring geographic dynamics of tuberculosis (TB) cases and incidences in Peninsular Malaysia from 2005 to 2014. In general, the TB distribution in the area seems to be in dynamic pattern and risk factors are also complex since the local clustering of the disease is not detected explicitly. The spatial pattern for the ten years TB cases shows an increasing trend, particularly in Selangor, Johor, and WPKL, but no definite clustering are determined in these states. Overall, an interesting hypothesis that could be further investigated is that why the majority of the cases occurred in the urbanised states such as Selangor, WPKL, and Johor, with high-medium incomes and populous areas. The findings are not only to answer the local TB dynamics and characterisation of high risk areas in general spatial perspective, but also provide a spatial knowledge for local researchers and decision makers to find, detect, and treat better solutions to the national TB strategies and challenges. Additional data on the risk factors and other spatial regression techniques can be further explored for more significant results.

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