Impacts of Lateral Boundary Condition Resolution in Tropical Urban Climate Modelling for Kuala Lumpur

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Abstract. Choosing the best LBCs is still debated among researchers due to the errors resulted. However, several recommendations have been documented to control the errors propagated by LBCs. One of the recommendations is employing higher resolutions LBCs. In the present, many LBCs are developed with various resolutions; spatially and temporally, for many applications but no claims regarding the best LBCs for tropical climate modelling have yet been documented. Therefore, this study intends to analyse the impacts of lateral boundary condition resolution during numerical downscaling within a tropical city. This study serves as a site-specific investigation to determine the suitable LBCs for the focused study area. Two widely used LBCs with different resolutions were utilized to initiate the Weather Research and Forecasting (WRF) simulation model. The performances of the two LBCs were compared using statistical tests and analyses. The study has found that the LBC with higher resolutions excels the other LBC during inter-monsoon season. Nevertheless, it was identified that both LBCs were able to provide reliable reconstruction of the tropical climate condition of the Kuala Lumpur City as portrayed by similar results obtained. Thus, it is concluded that both LBCs can be employed in numerical downscaling for tropical urban regions similar to the Kuala Lumpur City.

Keywords: Lateral boundary condition, resolution, WRF, tropical urban, Kuala Lumpur.

ENGINEERING JOURNAL Volume 25 Issue 1
Received 6 June 2020
Accepted 25 November 2020
Published 31 January 2021
Online at https://engj.org/
DOI:10.4186/ej.2021.25.1.165

This article is based on the presentation at the 4th International Conference on Research Methodology for Built Environment and Engineering 2019 (ICRMTEE 2019) in Bangkok, Thailand, 24th-25th April 2019.
1. Introduction

Obtaining the climate datasets is one of the main obstacles for studying climate condition. The evolution of technology has brought new approach of obtaining climate datasets which is through the implementation of numerical modelling [1]. Nowadays, general climate models (GCM) are used for climate studies worldwide. There are many GCMs developed by various institution to cater the needs which ensemble the climate circulation pattern via numerical modelling technique. However, the GCMs offer poor spatial resolutions for various scales of climate assessments especially for regional-scaled studies. The models succeed in resolving large-scale forcing but unable to represent the local climate impacts [2]. Due to this, regional climate models (RCM) are developed to simulate the GCMs using higher resolution nests over the targeted area [3]. This approach can account for the local climate impacts within the finer domains. In the urge for regional climate studies, numerical simulation models are often employed as a solution in obtaining climate datasets. These models are developed using complicated mathematical and physics theories (known as physics options) in order to resemble the complexity of climate.

Six physics options are included in these models namely microphysics, planetary boundary layer (PBL), long and short-wave radiation, surface layer, land surface layer and cumulus parameterization. Microphysics involves the theory of atomic, clouds and others to define the behaviour of the surrounding environment [4]. The PBL is included to set up the lowest atmospheric condition which directly influenced by the its contact to the planetary surface since this layer has rapid fluctuations of physical quantities such as temperature, moisture, flow velocity and others creating turbulence [5]. The influence of emissivity, absorption and surface fluxes within the atmosphere is modelled using long and shortwave radiation schemes [6]. Cumulus parameterisation has been developed over few decades to model the clouds within the atmosphere. The clouds are significantly influenced the climate system in the distribution of sensible and latent heat and momentum, reflection, absorption and emission of radiation, precipitation, and modification of radiation and planetary boundary layer processes [7]. In modelling regional climate, the interactions of the earth features towards the atmosphere are very crucial since they can alter the original concept of physics and dynamics in the atmosphere making every place on the Earth surfaces to be climatically unique [8]. Surface layer is included to solve the surface layer characteristics that includes turbulence, sensible heat flux, surface moisture flux and others [9].

Using the physics options, numerical models are used to downscale global-scaled weather datasets into regional scales which requires the usages of lateral boundary condition (LBC) for model initializations. In numerical models, the LBC utilized represents the Earth surfaces which is required to downscale the climate datasets from coarser scale into regional scales. In the case of nested models, the LBC is provided by the parent domain for its child. The errors produced by the LBC have been disputed by previous studies, however, recommendations were also suggested to minimize the errors [2], [10-12] including the selection of temporal and spatial resolutions. Davies [13] recently has also identified that errors produced is very small as compared to the overall errors. Though, the studies were performed for non-tropical regions, thus, least information on the tropical regions can be found [2], [10-13]. Based on this fact, site-specific study is encouraged to choose the best LBC for tropical climate modelling.

This study is conducted to analyse the performance of two different LBC in verifying the effects of urbanization on the thermal variation of a tropical urban climate like the Kuala Lumpur City. These two LBC were equipped with different resolutions are freely downloaded in the National Centre of Atmospheric Research (NCAR) archive. The result of this study can be be used as a guide for similar climate background. Future studies within the tropical regions are recommended to illuminate the errors resulted by applying the LBCs particularly when high resolutions climate modelling is required.

Over thousands of years, natural factors are the key drivers of climate changes such as volcanic activities and solar output variations [14]. However, in 1700s, urbanization was recognized as one of the significant factors of climate changes after the Industrial Revolution. The rate of the climate changes is also found to be at its highest peak in the 21st century due the rapid increment of human population. [14, 15]. In order to cater the human needs, many developments were erected and eventually turning many regions of the world into concrete jungles. It is indeed for human well-being as many job opportunities and better facilities are provided in the urbanized areas. The increment of the human population also resulted in the vast migration from rural to urban areas for better living [16]. Ironically, the urbanization has also turning the urban areas into heat islands when it becomes uncontrollable.

The urban thermal study has long been started [17-20]. Urban areas have been proven to degrade the climate condition in global and regional scales by elevating the ambient temperature by many studies [21-25]. The centre of the urban areas tends to have higher temperature compared to its outskirts. The intensity of the temperature decreases in areas further away from the centre of the urbanized area. This has been proven by many large-scaled studies on urban expansions. Most of these studies monitor the temporal and spatial changes of the urban areas as well as its relationships to its thermal environment. Recent studies have confirmed that the urban expansion; using spatial and temporal monitoring, contribute to the increase of the ambient temperature [22, 26-29]. Therefore, there is a need to investigate and scrutinize the reasons of these negative consequences behind the urbanization process.

Various studies on urban areas with more details have been conducted to identify the reasons behind the heating of the urban areas. Most of them suggest that the greatest
contributing factor of the urban heating is the increment of impervious surfaces by the man-made features such as buildings and pavements [17, 30]. The impervious surfaces tend to store more heat energy depending on their characteristics. The surface materials such as asphalt accounted as the contributor of heat energy by the impervious surfaces. Low albedo surfaces which fail to reflect the heat back on to the atmosphere during daytime makes the heat to be stored as latent heat by the urban features [31]. The heat will be released back into the atmosphere making the night time to be warm [32, 33]. This situation is worsened by limited sky view factor (SVF) which decreases the rate of heat release during night time [34].

High density of buildings is also one of the contributing factors to the degradation of climate in urban areas. Cities with high density of buildings will not permit the wind flows into the city which degrade the air ventilation system efficiency [35, 36]. The disruption of the wind will accumulate the heat and pollution within the cities since the strength of the wind is weaken by the building facades or surface roughness making the wind unable to sweep the heat and polluted air away from the city [37]. Other than that, high density buildings will limit the SVF and contribute to the night time warming as mentioned earlier. However, large SVF brings more heat during day time as the effects of shadow will be lessened [34]. The optimization of the SVF amount should be quantified for ideal heat release and shadow effects for positive impact towards the climate. Based on these, it is determined that urbanization is affecting the climate significantly. Since regional climate model is very crucial to study urbanization effects towards the climate, this study was also conducted to investigate the performance of the LBCs in determining effects of urbanization towards the variation of thermal environment.

2. Methods

In this study, a numerical model called WRF-ARW simulation model was employed to reproduce the air surface temperature of the study area. Remote sensing technique was used to extract the built-up area and GIS was utilized as the main analysis platform for the entire study. Parametric statistical methods were employed to analyse the performances of both LBCs. The methods employed for the study are discussed in the following sections.

2.1. WRF-ARW Simulation

In this study, the two global LBCs, NCEP FNL Operational Model for Global Tropospheric Analyses and NCEP GDAS/FNL for Global Tropospheric Analyses and Forecast Grids were chosen for comparison and denoted as Simulation 1 and Simulation 2 respectively. The characteristics of the LBCs utilized are presented in Table 1. These datasets were utilized due to their wide applications to resemble tropical conditions for regional climate downscaling [38-42]. The datasets were dated 17th April 2017 corresponding to the date of the Landsat satellite images employed during this study. The date chosen was within the inter-monsoon season phase when less synoptic forcing occurred as compared to monsoon seasons and another inter-monsoon season in October [43].

Table 1. LBCs' characteristics.

| Data Type       | Grid | Grid |
|-----------------|------|------|
| Data Format     | WMO_GRIB1 | WMO_GRIB2 |
| Temporal Resolution | 6 hours | 3 hours |
| Spatial (Grid) Resolution | 1˚ × 1˚ | 0.25˚ × 0.25˚ |
| Data Coverage   | From 0E to 359E and 90N to 90S | From 0E to 359E and 90N to 90S |

Fig. 1. The domains design and setup.

The regional modelling process was performed using WRF-ARW simulation to downscale the datasets. Domains with three nests were designed with the grid size of 37.5km (40 × 40), 12.5km (88 × 88), 2.5km (201 × 201) and 0.5km (151 × 151) to reproduce the near surface temperature. The grid dimensions were designed in easting and northing direction as shown in the parentheses. Figure 1 shows the domain designed for this study. Based on the figure, the four domains were designed using the same centre which focuses on the Kuala Lumpur location. This
is to provide relaxation zone for the domain edges to reduce the errors that might resulted by employing LBC [8]. The fourth domain (D04) covered the Klang Valley region where the study area is located, and the largest domain was designed to cover parts of Southeast Asia (D01).

Table 2. Physics schemes used during the simulation process.

| Physic Option            | Physic Scheme                                      |
|--------------------------|----------------------------------------------------|
| Microphysics             | WRF Single-Moment 3-class scheme                   |
| Longwave Radiation       | RRTM scheme                                        |
| Shortwave Radiation      | Dudhia scheme                                      |
| Surface Layer            | MM5 similarity                                     |
| Land Surface             | Noah Land Surface Model                            |
| Planetary Boundary Layer | Yonsei University                                  |
| Cumulus Parameterization | Kain-Fritsch                                      |

The WRF-ARW offers multiple physics options, simple as well as complicated schemes. Thus, a well-tried scheme combination was employed in configuring the physics and dynamics of the targeted area. The scheme combination used in the model as shown in Table 2. The study has employed the same domain design and scheme combinations for Simulation 1 and Simulation 2. The physics schemes chosen was a set of well-tried options conducted by [44] that successfully simulate the diurnal air surface temperature of the Kuala Lumpur city with the accuracy of over 90% agreements with the ground station observations provided by the Malaysian Meteorological Department (MMD). Using the same physics schemes as the study, the diurnal air surface temperature of the study area was simulated. The schemes have also been employed by many studies to remodel the actual urban climate condition within tropical countries [44-47].

2.2. Remotely-Sensed Data Extraction

This study has utilized the Landsat 8 Operational Land Imager (OLI) satellite image to extract the urbanized area. The image employed was dated 17th April 2017 during the first inter-monsoon season of the year. The image was also chosen due to a clear sky view that reduce the chance of atmospheric errors to occur [44]. The radiometric calibration procedure was conducted to correct the image to ensure it is free from radiometric errors [48]. The process converts the pixel values into surface reflectance as shown in Equation 1. Then, sun angle displacement was corrected using Equation 2.

\[
\rho_\lambda' = M_p \times Q_{CAL} + A_L
\]

where \( \rho_\lambda' \) = surface reflectance without sun angle correction

\[
\rho_\lambda = \frac{\rho_\lambda'}{\cos(\Theta_{SZ})} = \frac{\rho_\lambda'}{\sin(\Theta_{SE})}
\]

where \( \Theta_{SZ} = \) Local sun elevation angle
\( \Theta_{SE} = \) Local solar zenith angle

A combined algorithm developed by [49] was conducted to extract the built-up areas (urbanized areas). The algorithm is suitable to be used since the area of study share the same climate experiences as the Kuala Lumpur City. The study [49] suggested that the original NDBI algorithm developed derived by [50] should be corrected. Supported by [51]-[54], the confusion to separate the built-up areas from green cover and water features were suggested to be eliminated or reduced using Normalized Difference Vegetation Index (NDVI) and Modified Normalized Difference Water Index (MNDWI). Principal Component Analysis (PCA) is also suggested to be employed to assign the best pixel values for the MIR wavelength. The algorithm of NDBI, NDVI and MNDWI are shown as Equation 4, Equation 5 and Equation 6 respectively.

\[
\text{Built up Area} = \text{NDBI} - \text{NDVI} - \text{MNDWI}
\]

\[
\text{NDBI} = \frac{(\text{PCA Band 6.7} + \text{PCA Band 10.11}) - \text{Band 5}}{(\text{PCA Band 6.7} + \text{PCA Band 10.11}) + \text{Band 5}}
\]

\[
\text{NDVI} = \frac{\text{Band 5} - \text{Band 4}}{\text{Band 5} + \text{Band 4}}
\]

\[
\text{MNDWI} = \frac{\text{Band 3} - \text{Band 7}}{\text{Band 3} + \text{Band 7}}
\]
Subang station. The locations of these ground stations are presented in Fig. 2.

Fig. 1. Location of MMD ground stations.

2.4. Data Sampling and Inferential Statistics

Data filtration was conducted to ensure that the original data was free from outliers. As suggested by [55], [56], the process was carried out based on quartiles and boxplots to exclude the outliers. Prior to data analysis, the data was examined to identify the nature of the data distribution. The study has utilized the whole data samples and it is found that the data was approximately normal-distributed. This was indicated by the ratio of kurtosis and skewness value against the standard error. Thus, a series of parametric statistical approach was selected. Two statistical approach were used namely Pearson’s correlation test; to examine the interaction of urbanized areas with the air surface temperature, and one-way Analysis of Variance (ANOVA); to determine whether the impact is significant. The one-way ANOVA was conducted with the control of Welch’s and Brown-Forsythe’s test to ensure the error that may resulted can be reduced.

3. Result and analysis

The results and findings of this study are discussed into two parts; 1) the simulation performance and 2) the urbanization impacts towards the thermal environment of the Kuala Lumpur City. The results and findings are presented as follows:

3.1. Simulations Performance

The values of MAE, RMSE and R² for the two simulations are presented in Table 3. The results show that Simulation 2 excels Simulation 1 in reproducing the air surface temperature of the study area. It is based on the value of MAE of ±1°C, RMSE of ±1.4°C and R² of 0.881 compared to Simulation 1 which has slightly lower results; the value of MAE is ±1.5°C, RMSE is ±1.8°C and R² is 0.763. Figure 3 shows the diurnal pattern of the near-surface temperature of the city. Based on the figure, it is found that similar pattern was identified. Both simulations were found to underestimate the near-surface temperature in MARDI Serdang and Subang. However, both models were found to overestimate the near-surface profile during morning hours. On the other hand, both models tend to overestimate the temperature in FRIM Kepong despite some underestimation during the highest temperature readings. Through the study also, larger residuals were identified in Simulation 1 especially when the temperature was rapidly increase.

Table 3. MAE, RMSE and R² of Simulation 1 and Simulation 2.

|                  | NCEP FNL Operational Model Global Tropospheric Analyses, continuing from July 1999 (Simulation 1) | NCEP GDAS/FNL Global Tropospheric Analyses and Forecast Grids (Simulation 2) |
|------------------|------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------|
| MAE              | ±1.5°C                                                                                     | ±1.0°C                                                                         |
| RMSE             | ±1.8°C                                                                                     | ±1.4°C                                                                         |
| R²               | 0.763                                                                                      | 0.881                                                                          |

Fig. 2. Hourly profiles of near-surface temperature for both simulations in three different stations.
3.2. Urbanization Effects

3.2.1. Spatial distribution of urbanized areas and near-surface temperature profile

Figure 4 shows the distribution of built-up areas within the Kuala Lumpur City. It was found that, the city was dominated by built-up areas by approximately 85% of area coverage. The highest built area percentage found to be 99.5% indicating the imperviousness of the surface area. The lowest built-up percentage is 0% indicating there are no built-up coverage found within the area. Most of the highest built-up percentages are found in Sentul whereas the lowest built-up percentages were found in natural forest reserve such as Bukit Tabur, Bukit Besi and Bukit Gasing.

This study has identified that the northern, southern and central regions of the Kuala Lumpur City have higher built-up area percentages as compared to other regions. These areas are developed areas such as Sentul (residential), Kepong (industries), Central of Kuala Lumpur (business) and Bukit Bintang (business). These developed areas are expected to have higher near-surface temperature as compared to other less developed areas. Bukit Tunku and Mont Kiara are expected to have lower near-surface temperatures as the distribution of built-up areas are lower.

Spatial distribution of the daily mean air surface temperature of the Kuala Lumpur City produced by Simulation 1 and Simulation 2 are depicted in Figure 5. The lowest daily mean air surface temperature was 27.1°C and the highest value was 29.9°C. In Simulation 2, a close resemblance was identified. The lowest daily mean air surface temperature was 27.0°C and the highest mean temperature was 29.9°C. Based on Figure 5 also, the study has determined that the spatial distribution of both simulations was similar. In each simulation, the eastern regions own lower daily mean air surface temperature whereas the northern and southern regions experienced higher temperature variation.

In deeper examination of Figure 5, the lowest air surface temperature was found in the mountainous regions of Bukit Tabur which located in the north-easter region of the city. Mont Kiara, Bukit Tunku and Bukit Besi are other areas that own lower daily mean air surface temperatures. Sri Petaling has identified to have the highest daily mean air surface temperature which located in the south-western region of the city. Other regions with high daily mean air surface temperature are Kepong, Danau Kota, Bukit Bintang, Chow Kit, Kampung Baru and Segambut. Based on these results, it can be concluded that urbanized areas contribute to the increase of the air surface temperature within the Kuala Lumpur City.
3.2.2. Pearson’s correlation test result

The comparison made has determined that a close result was obtained for Simulation 1 and Simulation 2. Based on the Pearson’s correlation test, the correlation between the daily mean air surface temperature and built-up percentage for Simulation 1 and Simulation 2 were 0.662 and 0.648 respectively with a positive relationship. The relationship was indicated by the increasing trendline as portrayed in Figure 6. As shown in the figure, the distribution of the points in the graph is closer in Simulation 2 which explained the slightly higher result. Thus, in presenting the correlation between the daily mean air surface temperature and built-up area percentage, both simulations gave a very similar result. Therefore, both LBC can be employed for similar climate background studies.

![Fig. 5. Trendline of near-surface temperature and urbanized area for Simulation 1 and Simulation 2.](image)

3.2.3. One-way ANOVA test result

The investigation of the performance of both LBCs employed was further by conducting one-way ANOVA procedures. A similar test was performed for both simulations. Using the daily mean air surface temperature produced, this study has identified that both models a similar result; where both test rejecting the null hypothesis which suggesting the same conclusion. The effects of urbanization towards the near-surface temperature were identified as significant as the p-value rejected the null hypothesis of one-way ANOVA. Through the test, it was confirmed that there was at least one significant difference between the classes of data samples. This result suggests that the changes in the built-up area percentage will significantly affect the near-surface temperature value within the Kuala Lumpur City.

Previously, many studies have documented that rapid urbanization leads to fast conversion of green covers such as natural forest and croplands which is the prime cause to the rise in the surrounding temperature that leads to the formation of UHI [25], [26]. The green covers and croplands are replaced with built-up areas which increase the heat capacity storage. This will lead to the increase of heat release into the air, making the surrounding temperature to be hotter. The present study also discovered similar phenomena occurred in the Kuala Lumpur City as documented by previous studies. The region with high built-up areas experienced high temperature as compared to the area with less built-up areas [40, 43].

4. Conclusion

The performance of two LBCs with different spatial and temporal resolutions in initializing the boundary conditions in WRF-ARW numerical simulation were analysed in this study. The performance was tested in regenerating the near-surface temperature of a tropical city in Malaysia. The study has identified that the LBC with higher spatial and temporal resolution gave higher performance for regional downscaling. However, this study has also identified that both LBCs offers similar performance, thus, both LBCs are suitable to be utilized in reproducing the near-surface temperature of tropical cities. Future studies regarding this matter are urged with separate analysis on spatial and temporal resolutions impacts in regenerating the tropical urban climate condition.

As Malaysia envisions its cities to be climatic-friendly in the near future, implementation of the urban climatic aspect should be emphasized in urban planning process. In line with the National Policy on Climate Change and the National Green Technology Policy introduced by the government, consideration on the urban climatic aspect
require active participation from the multi-disciplines experts. At the moment, international efforts in addressing the importance of sustainable development are also evident through many programs such as the Global Cool Cities Alliance (GCCA), World Urban Forum (WUF) and New Urban Agenda by United Nations which encourage the countries worldwide to join the alliance to combat the climate change impacts on urban regions.

Acknowledgements

The authors would like to express their gratitude to the Universiti Teknologi MARA (UiTM) for funding this project under Geran Bestari Perdana 600-IRMI/DANA 5/3 BESTARI (P) (074/2018).

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