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**Two Decades of Endemic Dengue in Bangladesh (2000-2022): Trends, Seasonality, and Impact of Temperature and Rainfall Patterns on Transmission Dynamics**

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**Abstract:**

**Background:** The objectives of this study were to compare the dengue virus (DENV) infection, deaths, case-fatality ratio, and meteorological parameters between the first and the recent decade (2000-2010 vs. 2011-2022) and to understand the trends, seasonality, and impact of change of temperature and rainfall pattern on transmission dynamics of dengue in Bangladesh

**Methods:** For the period 2000-2022, dengue cases and death data from Bangladesh’s Ministry of Health and Family Welfare’s website, and meteorological data from the Bangladesh Meteorological Department were analyzed. A Poisson regression model was performed to identify the impact of meteorological parameters on the monthly dengue incidence. A forecast of dengue cases was performed using an autoregressive integrated moving average model.

**Results:** Over the past 22 years, a total of 244,246 dengue cases were reported including 849 deaths (Case fatality ratio [CFR] =0.34%). The mean annual number of dengue cases increased eight-fold during the second decade, with 2216 cases during 2000-2010 vs. 18,321 cases during 2011-2022. The mean annual deaths have doubled (21 vs. 46), but the overall CFR has decreased to one-third (0.69% vs 0.23%). The annual temperature increased by 0.49 °C, and rainfall decreased by 314 mm. Monthly mean temperature (Incidence risk ratio [IRR]: 1.26), first-lagged rainfall (IRR: 1.08), and second-lagged rainfall (IRR: 1.17) were significantly associated with monthly dengue incidence.

**Conclusions:** The increased local temperature and unusual rainfall might have contributed to the increased incidence of DENV infection in Bangladesh. Community engagement, vector control, and destruction of mosquito habitats are key to controlling dengue.

**Keywords:** Dengue, Bangladesh, Climate change, Temperature, Rainfall

**Introduction:**

Dengue fever is a mosquito-borne disease (MVD) caused by four distinct serotypes of the dengue virus (DENV) of the Flaviviridae family (WHO 2009). DENV is transmitted to humans by the bites of the female Aedes species mosquitoes including *Ae. aegypti* and *Ae. albopictus* (WHO 2009, CDC 2019). DENV is endemic in over 125 countries of the world and the number of cases globally reported to WHO continues to increase every year (WHO 2023a). Annually, an estimated 390 million dengue infections are recorded across the world, including 96 million clinical cases (Murray et al. 2013, WHO 2023b). Most infections (>80%) are self-limiting with no or mild clinical manifestation resulting in lifelong immunity for serotype (WHO-Bangladesh 2022). However, infections with different serotypes, known as secondary dengue infection, may result in severe dengue with a higher case-fatality ratio (Teo et al. 2023).

Currently, South and Southeast Asia is considered to be the hotspot of DENV infection with more than 50% of cases recorded in the regions (WHO South-East Asia 2023). The first official DENV outbreak in Bangladesh was reported in 2000, and since then, dengue has become endemic in the country posing a significant health challenge (Sharmin et al. 2015). Over the past few years, the number of dengue cases has been steadily increasing with significant seasonal and regional variations. Analysis of data from 2000 to 2017 revealed that almost half of the dengue cases occurred during the monsoon season (May-August) and the post-monsoon season (September-December) (Mutsuddy et al. 2019). However, a shift in seasonal patterns has been observed since 2014, with dengue cases being reported during the pre-monsoon season as well (Mutsuddy et al. 2019). During 2015-2017, the number of dengue cases during the pre-monsoon season was more than seven times higher compared to the previous 14 years (Mutsuddy et al. 2019).

Climate change including changes in precipitation, temperature, and humidity, as well as rapid unplanned urbanization, were identified as strong predictors of an ecological imbalance that has led to an increase in dengue cases in Bangladesh (Mutsuddy et al. 2019). This suggests that the dengue transmission season could eventually extend year-round, with a higher chance of outbreaks occurring at any time of the year. Identifying trends and seasonality in dengue cases can aid health authorities and relevant public and private administrations in effectively allocating resources to control the spread of the DENV through vector control. The objectives of our study were to: i) compare the annual and monthly cases in the first [2000-2010] and recent decade [2011-2022], ii) identify the trend and seasonality of dengue cases, iii) quantify the impact of climatic parameters for the monthly incidence of dengue cases in the country and iv) forecast the annual incidence of dengue cases for next decade.

**Materials and Methods:**

**Data sources:**

The data on the number of reported dengue‐infected people have been extracted from the Directorate General of Health Services (DGHS)'s website from January 2000 to December 2022 (DGHS 2023). We used the definition of dengue cases used by the Ministry of Health and Family Welfare, Bangladesh, which was discussed in our earlier article (Ahsan et al. 2021). We collected three-hourly temperature and daily rainfall data from Bangladesh Meteorological Department (BMD) over the period 2000–2022 (BMD 2023) for the meteorological station located in Mirpur, Dhaka.

**Variables**

The monthly number of dengue cases was used as the primary outcome variable. Two climatic variables- temperature and rainfall are used as the covariates for the regression analysis. In addition, two lagged variables rainfall in lag 1 and lag 2 have also been used as predictors for the incidence of monthly dengue cases to capture the actual impact of those meteorological parameters. We also used monthly mortality data for comparison between two decades.

**Statistical analysis**

We analyzed the monthly dengue incidence and meteorological data for the period of 2000-2022. In the first stage, descriptive analysis was conducted to determine the characteristics of confirmed dengue cases and deaths with mean, and standard deviation in each year and each month for the entire period. Then, we compared dengue cases, deaths, and weather parameters in two decades (2000-2010 and 2011-2022) using paired sample t-test. Next, we calculated the monthly growth factor (GF) of dengue cases by dividing the number of dengue cases reported in each month by the number of dengue cases reported in the previous month and repeating this process for each month from 2000 to 2022 (Haider et al. 2021). The formula for the growth factor can be given by

where indicates the number of dengue cases in th month. To avoid the occurrence of zeros in some months, we added 1 to the total number of cases for each month. This allows us to obtain a real-valued measurement of the GF for the above equation. The distribution of GF was skewed; therefore, we used the first natural log transformation before the data was further examined. However, we have also performed a reverse transformation of the log (GF) values by exponentiating values to convert them to the original scale for ease of interpretation(Haider et al. 2021).

We performed forecasting using the autoregressive integrated moving average (ARIMA) model. The ARIMA model is a data-driven, exploratory strategy that enables us to fit a suitable model and forecast values. The ARIMA model consists of autoregressive (p) terms, differencing (d) terms, and moving average (q) operations, and it is denoted as ARIMA (p, d, q) (Kumar and Susan 2020). To select the appropriate autoregressive and moving average orders, the autocorrelation function (ACF) and partial autocorrelation function (PACF) were examined. Additionally, the differencing parameter, represented by "d," indicates the number of times the time series is different to achieve stationarity. An ARIMA (p, d, q) process refers to an autoregressive moving average (ARMA) model that has been differenced "d" times to obtain stationarity (Hasan et al. 2021). By removing high-frequency noise from the data, the model discovers local patterns by assuming that the time series values are linearly related. We also conducted a Mann-Kendall (M-K) trend analysis to determine possible upward or downward trends (Yue and Pilon 2004). We also performed the Sen's slope test to assess variations in annual dengue cases and deaths (Sen 1968).

We, then used a time series count generalized linear model (GLM), more specifically, a time-series Poisson regression model to determine whether the climatic factors were associated with the dengue cases over time (Sumi et al. 2021). The non-normality, heteroscedasticity, and non-linearity that characterize count data can be fitted easily using GLMs. The time-series observations may possess autocorrelation and they might be nonnegative integers, and thus GLM is useful in overcoming both issues [20, 21, 22]. Monthly dengue cases were utilized as the outcome variable in this model, along with data from the Bangladesh Meteorological Department (BMD) on temperature and rainfall. To capture the actual impact of rainfall on dengue incidence across time, we additionally employed two lagged variables of meteorological elements, mainly rainfall in lag 1 and 2. After eliminating predictors with a higher multicollinear relationship, we have arrived at average temperature, rainfall (in lag 1), and rainfall (in lag 2) as the final set of predictors for the monthly dengue incidence in Bangladesh. We used the statistical program RStudio, version 3.5.2.2 for the analyses (R Core Team 2022).

**Results:**

Between 2000 and 2022, Bangladesh reported a total of 244,246 dengue cases with an annual mean of 10,161 cases (±standard deviation [SD]=23,971) including 849 fatal outcomes indicating a case-fatality ratio (CFR) of 0.34%. Between 2000 to 2010, the mean annual number of dengue cases was 2,216 (±2,123) which has increased over eight folds compared to the following decade (2011-2022) at 18,321 (±31,778) **(Table 1)**. Between these two periods, the mean number of annual deaths due to DENV infection has increased by 2.2 times (21.18 cases vs 46.58 cases). However, the CFR of DENV infection has decreased to almost one-third between two decades (0.69 vs 0.23) **(Table 1)**.

The highest monthly average number of cases was recorded in August (n=3407 cases) and the lowest was in March (n=6.7 cases) **(Fig 1)**. The highest number of annual cases was reported in 2019 with 101,354 and the highest number of deaths was recorded in 2022 with 281 deaths, which was 35% of total deaths recorded in the past 23 years in Bangladesh **(Fig 1)**. Most of the dengue-related deaths were recorded after 2018, with more than 69% (n=550) deaths recorded during this time **(Fig 1)**.

The average annual temperature was 26.35°C (SD=0.49) during the first decade (2000-2010) and 26.84°C (SD=) during the recent decade (2011-2022) (**Table 1**). The increase of 0.49°C temperature was equivalent to 4292 degree-hour/year of heat (365 days X 24 hours X 0.49°C). The annual rainfall has decreased by 314 mm between two decades (2078.66 mm vs. 1764.50 mm) (**Table 1**), of which 308 mm decreased during the monsoon (July-October) season and only 6 mm decreased during the non-monsoon period. Compared to the first decade (2000-2010), an unusually higher amount of monthly precipitation has been observed in the second decade (2011-2022) with most of the months recording extreme rainfall (more than 3rd quantile value of monthly rainfall for the decade) shown as an outlier of the box plot **(Fig 2)**.

The overall mean GF from month to month was 1.37 (SD=0.86). However, in four months (April-July), the monthly GF was above one (lower 95% confidence interval >1), while for the rest of the years, the monthly GF was less than 1 (95% confidence interval crossed 1). More than 77% (71/92) of months between April and July for the period 2000–2022 had mean monthly GF > 1 compared to only 16% (30/184) of months between August and March of the same period. June had the highest GF with a mean value of 3.47 indicating that cases would be more than three times higher in the next month (July). The lowest GF was recorded in December with a mean of 0.54 (95% CI: 0.40 to 0.69) indicating that cases in January would be halved compared to the number of cases recorded in December (**Fig. 3**). In the M-K trend analysis, we found a positive trend of reported dengue cases (p <0.001 and tau = 0.26). In Sen’s slope test, the slope was 171.67 (95% CI: -46 to 687) indicating an upward trend in upcoming months (**Table 2**).

In the GLM, the estimated effect of each variable is presented as the incidence risk ratio (IRR). The model suggests that dengue cases would rise by 26% for a one-degree centigrade (°C) temperature increase. For each additional centimeter (cm) of rainfall in the first lagged month, the number of dengue cases increased by 8% (IRR= 1.08 [95% CI: 1.08-1.09]), and in the second lagged month increase the cases by 17% [IRR=1. 17 (95% CI: 1. 17 -1.18)] **(Table 3)**.

In the ARIMA model, we detected an increasing trend for the first few years, which then started to decline. However, a stiff rise in cases was observed after 2018 except for 2020 (the first year of the Covid-19 pandemic). The forecasted value showed a continuously increasing trend of DENV infection in Bangladesh **(Fig 4)**.

**Discussion:**

Dengue is currently a worrying and important public health challenge for Bangladesh. Our analysis showed that the number of dengue cases has increased eight times and deaths have doubled, and the CFR dropped to one-third between the first and second decade of this century in Bangladesh. Between these periods, the annual temperature increased by 0.49°C, and annual rainfall decreased by 314 mm, despite changes in rainfall patterns with unusually early or late rainfall outside the typical monsoon season in Bangladesh (July-October) (Haider et al. 2014). The monthly growth factor remains above one significantly for four months (April to July) which overlaps the hot and humid period of the year. Monthly mean temperature, monthly first-lagged rainfall, and second-lagged rainfall played a critical role in monthly dengue incidence in Bangladesh.

The increase of 0.49°C temperature adds approximately 4292-degree-hours equivalent heat per year in the country. This additional heat would favor mosquito-borne disease transmission. For dengue virus transmission, approximately 305-degree-hours equivalent heat is needed to accomplish the extrinsic incubation period in Aedes mosquitoes at 26°C (Focks et al. 1995). Thus, the additional 0.49°C temperature will add the burden of more than 14 generations of infectious mosquitoes in the environment of Bangladesh. An 8-fold increase in dengue cases is an indication of such changes in temperature in the country. Our model identified a significant role of monthly mean temperature with an additional 1°C temperature increasing the monthly cases by 26%. Earlier studies showed that for every 1°C increase in temperature, dengue cases increased by 61% in Australia, 12-22% in Cambodia, 5% in Vietnam, and 2.6% in Mexico (Soneja et al. 2021).

Rainfall facilitates mosquito breeding and plays an important role in mosquito-borne disease transmission. Although we found a 15% reduction in annual rainfall in the recent decade from the immediate past decade, we found an increase in unusually high rainfall in pre-and-post monsoon season. Our model showed that both the first and the second lagged month’s rainfall increased monthly cases by 8% and 17%, respectively. These findings are consistent with earlier studies in Bangladesh that showed that peak dengue cases occurred two months after the peak rainfall (Salje et al. 2016) or an additional rainy day per month increased dengue cases by 6% in the succeeding month (Rahman et al. 2020). Similar findings were reported in Vietnam with the dengue incidence being associated with both first and second-lagged months (Cuong et al. 2011). In Timor-Leste, a 47% increase in dengue incidence was recorded with an additional 1 mm seasonal rainfall increase (Wangdi et al. 2018). These findings are biologically plausible as rainfall allows approximately two generations of dengue cases over a month. A generation interval is a time difference between a primary human infection and a second human infection originating from the first human case through two bites of mosquitoes (Siraj et al. 2017). To accomplish a generation interval the virus and mosquito undergo several phases including intrinsic incubation period in humans, human-mosquito transmission (first bite), extrinsic incubation period in mosquitoes, blood meal digestion period, and finally mosquitoes-to-human transmission (2nd bite) (Siraj et al. 2017). Ideally, for DENV, the generation interval completes at around 16 days at 28-32 °C (Siraj et al. 2017).

Bangladesh’s dengue season is characterized by hot and wet periods running between June to August. This is the period with the highest amount of rainfall in the country facilitating Aedes mosquito breeding in the country (Haider et al. 2023). The monthly mean growth factor above 1 for April – June indicates that for each of these months, the incidence of dengue cases will surpass the current month. Thus, we suggest starting vector control intervention in April in Bangladesh.

Globally and regionally in South and Southeast Asia, dengue cases are increasing. DENV infection increased by more than 46% between 2015 and 2019 in the region (WHO South-East Asia 2023). In 2023, up until 31 May, a total of 1515,460 DENV infections were recorded in Brazil with 387 deaths (European CDC 2023). In Malaysia, a total of 43,619 DENV infections have been recorded by 21 May 2023 (European CDC 2023). We found an increasing trend of DENV infection in Bangladesh. This increasing trend was much stiffer after the serotype DENV-3 was introduced in the country in 2018 (Ahsan et al. 2021). This increased trend is possibly linked with climate change in the region attributed to increased temperature and unusual rainfall, urbanization, population growth, inadequate water supply and storage practice, poor sewer, and waste management system, rise in global commerce and tourism (WHO South-East Asia 2023).

The case fatality ratio (CFR) of primary dengue infection is very low with an estimation of 0.018% - 0.1% (Huits and Schwartz 2021). However, the CFR of secondary dengue infection is high, although precise estimates are not available, some studies show more than 1% and reaching up to 4% (Liu et al. 2020). Bangladesh’s overall CFR of dengue infection (0.34%) seems slightly higher considering the overall CFR reported in other South and Southeast Asian countries (WHO South-East Asia 2023). However, more than 69% of dengue-related deaths in Bangladesh were recorded after the introduction of the serotype DENV-3 in 2019. Thus, secondary infection is likely contributing to higher dengue-related deaths in Bangladesh. In addition, the CFR of the dengue virus infection might have been affected by a lack of active surveillance and missing the mild and asymptomatic cases, and not recording the cases outside the public hospital and few selected private hospitals in Bangladesh or weaker health care system in the country (Ahsan et al. 2021). In some years, the CFR was very high, for example, in the year 2003, the CFR was 2.1 (total cases 486), in the year 2000, 1.68 (total cases 5,551), and in 2022, 0.45 (total cases 62,382). On the other hand, the CFR decreased in the second decade. This improvement is probably associated with improved access to the health care system, a better understanding of the treatment protocol including the availability of clinical management guidelines and training for the health care providers, better availability of Information, Education, and Communication (IEC) materials, community engagement and expansion of surveillance system to more hospitals in the surveillance system across the county in the recent years, and overall improvement of the economic condition of the country (Diseases Control Division (DGHS) 2013, WHO 2017, Albis et al. 2019).

Two large dengue outbreaks occurred in Bangladesh in the year 2019 and 2022 both characterized by unusual weather patterns and the occurrence of two different serotypes. The 2019 outbreak was characterized by early rainfall of 120 mm in February compared to a monthly mean of 20 mm precipitation, along with the introduction of a new serotype of DENV-3 in the country (Ahsan et al. 2021). The 2022 outbreak was characterized by the late onset of rainfall with 297 mm rainfall in October compared to a monthly mean of 156 mm, and thus prolongation of vector transmission season along with the introduction of a new serotype, DENV-4 in the country (Haider et al. 2023). The occurrence of a new serotype exposed a large naïve population in a densely populated country like Bangladesh. A large proportion of the population is already infected with one of the serotypes of DENV with more than 80% of people living in Dhaka having antibodies against DENV (Salje et al. 2016). Another study predicted an estimated 40 million people being infected with DENV nationally and 2.4 million annual infections (Salje et al. 2019a). Thus, any subsequent infections raise the risk of developing severe dengue hemorrhagic fever through antibody-dependent enhancement (ADE) (Teo et al. 2023). The deaths of many people in the year 2022 when the new serotype DENV-4 was introduced were probably associated with secondary dengue infection.

Controlling vector-borne diseases in tropical countries where temperatures, humidity, and rainfall remain favorable for breeding mosquitoes during most periods of the year is a difficult task (Haider et al. 2023). Concerns were raised over the development of insecticide resistance (Al-Amin et al. 2020, Ahsan et al. 2021) and the failure of developing a successful dengue vaccine (Wang et al. 2017). The prospect of *Wolbachia-*related intervention is bright but still far from applying on a national scale considering the expenses and technicalities associated with this. In this situation, an integrated and holistic vector management plan while engaging the local communities is key for controlling Aedes-borne diseases, especially in resource-limited countries. Regular destruction of mosquito breeding sites and increasing surveillance for detecting active cases are key in controlling dengue virus infection. Continuous active dengue surveillance will enable early detection of cases and outbreaks. Public health authorities will be able to identify areas where the disease is spreading, take immediate action to control mosquito populations, isolate infected patients, and implement public awareness campaigns to educate people about preventive measures. Early detection and response can help prevent the further spread of the disease and reduce its impact on individuals and communities.

Regular destruction of mosquito breeding habitats and increasing surveillance for detecting active cases should prioritize in controlling dengue virus infection in Bangladesh. Policymakers need to design an Aedes-borne disease management plan by considering a range of diseases that Aedes mosquito can transmit including Chikungunya, yellow fever, Zika virus, West Nile, Japanese Encephalitis, Eastern Equine Encephalitis, Ross River, Rift Valley fever, and the LaCrosse virus (Haider et al. 2023).

Our data should be viewed instead of several weaknesses of our study. We relied on the reported number of cases from the Ministry of Health and Family Welfare’s website, which mainly relies on passive reporting systems from the selected health facilities in the country (Ahsan et al. 2021). These numbers seem to be an underestimation of actual cases. A modeling study based on the national seroprevalence of DENV antibodies predicted an annual infection of 2.4 million cases (Salje et al. 2019b). However, dengue infection is underestimated globally as it is difficult to diagnose asymptomatic or mild cases that never reach healthcare settings. Although mild cases are missed more frequently, the severe and fatal cases would likely visit the hospital and thus be counted as numerators in our estimation. Thus, our estimation did not overlook the worst-case scenario, that is, our estimation, for example, estimated the higher CFR rather than the lower possible estimates.

**Conclusions:**

Between the first (2000-2010) and the second decade (2011-2022), dengue cases have increased by 8.3 times, and annual deaths have increased by 2.2 times in Bangladesh. This growth of cases is partly explained by the influence of global warming with an increase of 0.49°C annual temperature as well as changes in duration and length of the rainy season. Unusual rain including early or late rain in and beyond the monsoon season likely contributed to extending the length of the dengue transmission season in Bangladesh. The monthly mean temperature, and monthly total rainfall of the first-lagged month and second-lagged months showed a greater influence on the monthly incidence of DENV infection in Bangladesh. The mean monthly growth factor remains significantly above one during April-July, which coincides with the hot and rainy season of the country indicating an earlier vector control would benefit the country. The ARIMA model forecasts a continuously increasing trend of DENV infection for the next decade in Bangladesh. We recommend an integrated and holistic vector management plan while engaging the local communities in the regular destruction of mosquito breeding sites and increasing surveillance for detecting active DENV-infected cases. Proactive surveillance, vector control, and vaccine rollout remain essential public health interventions. In the context of climate change, urbanization, trade, and the movement of people with vectors, there is a need to operationalize the One Health approach to address dengue fever and other vector-borne diseases in Bangladesh and beyond.

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**Author contribution statement:** NH ideated the study and all authors helped develop the study outline and protocol. MNH and IK collected the data. NH, MNH, MA, and AZ analyzed the data. NH, IK, and MNH prepared the first draft manuscript and all authors contributed to several drafts and finalization of the manuscript. All authors approved the final draft and submission of the manuscript.

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**Conflict of interest:** The authors declare that they have no conflict of interest.

**Ethics statement:** This study does not include any individual-level data and thus does not require any ethical approval. We used publicly available data on Dengue cases and deaths.

**Data availability statement:** All the dengue data presented in this manuscript are publicly available on Bangladesh’s Ministry of Health and Family Welfare’s Directorate General of Health Services website (<https://dghs.gov.bd/> ). The meteorological data were purchased from Bangladesh Meteorological Department and are restricted to use for research purposes only and anyone interested in these data can request Bangladesh Meteorological Department (<https://live3.bmd.gov.bd/> ).

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**Tables and Figure Legends:**

**Tables:**

**Table 1: Comparison of dengue cases, deaths, and weather parameters between the first (2000-20210) and the recent decade (2011-2022) in Bangladesh**

**Table 2: The Mann-Kendell trend test for reported dengue cases in Bangladesh, 2000-2022**

**Table 3: The incidence risk ratio (IRR) of monthly average temperature and total rainfall for monthly incidence of Dengue cases in Bangladesh using time-series count Generalized Linear Model for the period 2000-2022**

**Figures:**

**Fig 1:** Top: Number of dengue cases and deaths over the period 2000-2022, Bangladesh. Bottom: Number of monthly dengue cases and deaths recorded in Bangladesh.

**Fig 2:** The boxplot compares monthly rainfall in Dhaka city, Bangladesh between two decades (2000-2010 vs 2011-2022). The bottom and top of the box indicate the first and third quantiles, the band inside the box is the median. The dots outside the box are individual outliers. Most of the months in the second decade had outlier rainfall whereas in the first decade, only the cooler months (Nov-Jan) had some extreme rainfall.

**Fig 3: Top:** Mean monthly growth factor for the period of 2000-2022. **Bottom:** The Monthly growth factor for the individual year 2000-2022. The dotted horizontal line indicates monthly growth factor 1 (same number of cases in two subsequent months).

**Fig 4:** The observed and forecasted number of dengue cases in Bangladesh using the Autoregressive moving average (ARIMA) model including a 95% confidence interval.