

Is heat wave a predictor of diarrhoea in Dhaka, Bangladesh? A time-series analysis in a South Asian tropical monsoon climate --Manuscript Draft--

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Corresponding Author:	Farhana Haque, MBBS MPH MSc PhD LSHTM: London School of Hygiene & Tropical Medicine LONDON, UNITED KINGDOM
Order of Authors:	Farhana Haque, MBBS MPH MSc PhD Fiona C Lampe, BSc MSc PhD Shakoor Hajat, BSc MSc PhD Katerina Stavrianaki, PhD SM Tafsir Hasan, MBBS MS ASG Faruque, MBBS MPH Tahmeed Ahmed, MBBS PhD Shamim Jubayer, BDS MPH DYDW MS Ilan Kelman, BSc MSc PhD
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Abstract:	While numerous studies have assessed the association between temperature and diarrhoea in various locations, evidence of relationship between heat wave and diarrhoea is scarce. We defined elevated daily mean and maximum temperature over the 95th and 99th percentiles lasting for at least one day between March to October 1981 – 2010 as TAV95 and TAV99 and D95 and D99 heat wave, respectively. We investigated the association between heat wave and daily counts of hospitalisations for all-cause diarrhoea in Dhaka, Bangladesh using time series regression analysis employing constrained distributed lag-linear models. Effects were assessed for all ages and children aged under 5 years of age. Diarrhoea hospitalisation increased by 6.7% (95% CI: 4.6% – 8.9%), 8.3% (3.7 – 13.1), 7.0 (4.8 – 9.3) and 7.4 (3.1 – 11.9) in all ages on a TAV95, TAV99, D95 and D99 heat wave day, respectively. These effects were more pronounced for under-5 children with an increase of 13.9% (95% CI: 8.3 – 19.9), 24.2% (11.3 – 38.7), 17.0 (11.0 – 23.5) and 19.5 (7.7 – 32.6) in diarrhoea hospitalisations on a TAV95, TAV99, D95 and D99 heat wave day, respectively. At lags of 3 days, we noticed a negative association indicating a 'harvesting' effect. Given that no heat wave definitions exist, and no heat warnings are issued at present, these results may help to define heat waves for Dhaka and trigger public health interventions including heat alerts to prevent heat-related morbidity in Dhaka, Bangladesh.
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Is heat wave a predictor of diarrhoea in Dhaka, Bangladesh? A time-series analysis in a South Asian tropical monsoon climate

Farhana Haque^{1,2*}, Fiona C Lampe¹, Shakoor Hajat³, Katerina Stavrianaki⁴, S. M. Tafsir Hasan⁵, ASG Faruque⁵, Tahmeed Ahmed⁵, Shamim Jubayer⁶, Ilan Kelman^{7,8}

¹Institute for Global Health (IGH), University College London (UCL), London, UK

²UK Public Health Rapid Support Team (UK-PHRST), Department of Infectious Disease Epidemiology, London School of Hygiene and Tropical Medicine (LSHTM), London, UK

³Department of Public Health, Environments and Society, London School of Hygiene and Tropical Medicine (LSHTM), London, UK

⁴Institute for Risk and Disaster Reduction (IRDR), University College London (UCL), London, UK

⁵Nutrition and Clinical Services Division, icddr,b, Dhaka, Bangladesh

⁶National Heart Foundation Hospital and Research Institute (NHF&RI), Dhaka, Bangladesh

⁷Institute for Global Health (IGH) and Institute for Risk and Disaster Reduction (IRDR), University College London (UCL), London, UK

⁸University of Agder, Kristiansand, Norway

22 ***Corresponding author:**

23 Farhana Haque

24 Email address: farhana.haque@lshtm.ac.uk; farhana.haque@alumni.ucl.ac.uk

25 Current postal Address: Room 118, Keppel Street, London, WC1E 7HT, UK

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Abstract

While numerous studies have assessed the association between temperature and diarrhoea in various locations, evidence of relationship between heat wave and diarrhoea is scarce. We defined elevated daily mean and maximum temperature over the 95th and 99th percentiles lasting for at least one day between March to October 1981–2010 as TAV95 and TAV99 and D95 and D99 heat wave, respectively. We investigated the association between heat wave and daily counts of hospitalisations for all-cause diarrhoea in Dhaka, Bangladesh using time series regression analysis employing constrained distributed lag-linear models. Effects were assessed for all ages and children aged under 5 years of age. Diarrhoea hospitalisation increased by 6.7% (95% CI: 4.6% – 8.9%), 8.3% (3.7 – 13.1), 7.0 (4.8 – 9.3) and 7.4 (3.1 – 11.9) in all ages on a TAV95, TAV99, D95 and D99 heat wave day, respectively. These effects were more pronounced for under-5 children with an increase of 13.9% (95% CI: 8.3 – 19.9), 24.2% (11.3 – 38.7), 17.0 (11.0 – 23.5) and 19.5 (7.7 – 32.6) in diarrhoea hospitalisations on a TAV95, TAV99, D95 and D99 heat wave day, respectively. At lags of 3 days, we noticed a negative association indicating a ‘harvesting’ effect. Given that no heat wave definitions exist, and no heat warnings are issued at present, these results may help to define heat waves for Dhaka and trigger public health interventions including heat alerts to prevent heat-related morbidity in Dhaka, Bangladesh.

Keywords

heat wave, diarrhoea, heat effects, extreme heat, temperature extreme, Bangladesh

49 Introduction

50 It is now evident that anthropogenic climate change is increasing the intensity
51 and frequency as well as duration of heat waves in addition to raising the average
52 ambient temperature across the globe (1, 2). The observed increasing trend of heat
53 waves and warm spells due to global climate change are projected to continue in the
54 future (3, 4). Heat waves can exert serious and potentially life-threatening impacts on
55 human health including heat stroke, heat exhaustion, heat syncope, and heat
56 cramps (5). Heat extremes have been associated with excess all-cause mortality,
57 increased emergency room visits and hospital admissions, increased mortality from
58 cardiovascular and other diseases, mental health issues, adverse pregnancy and
59 birth outcomes and increased healthcare costs (5-12).

60 Health effects of heat wave tend to be governed by a variety of complex,
61 interacting biological, medical, environmental, social and geographical factors
62 including locations, individual susceptibility, prevalence of certain diseases,
63 healthcare infrastructure and health system status (2, 5, 7, 13). In addition, the
64 mechanisms by which extreme temperatures influence disease causation may vary
65 widely according to different morbidities. For example, heat extremes in countries
66 with less than optimum water and sanitation infrastructure may significantly increase
67 the risk of waterborne diseases including diarrhoea by increasing exposure to
68 contaminated drinking water needed to replace the volume lost through excessive
69 sweat in addition to increasing host susceptibility to infection (5).

70 Although there have been several reports of increased mortality, limited
71 information exists on the impact of heat waves on morbidity across the globe and

particularly in the South Asian context (14, 15). The perceived risk of health hazards from heat waves or warm spells is low in the developing countries of the tropical and sub-tropical regions in South Asia where comfortable warm temperature is the norm (15). Although temperature-related deaths and diseases may be largely preventable and heat warning systems (HWSs) as well as heat early warning systems (HEWS) are existent in many high-income cities globally, such warning systems rarely exists in the South Asian setting (11). One important gap that hampered the development of a warning system in South Asian countries is the lack of consensus about the definition of heat waves. Furthermore, there is dearth of knowledge regarding the nature of heat-health risk, climate hazard, societal exposure and population vulnerability (15).

Bangladesh, a South Asian country with a tropical monsoon climate, is highly vulnerable to the adverse impacts of climate change (16) and heat waves in the future (15). With more than an estimated 76 million people affected with diarrhoeal disease episodes in all age groups in Bangladesh annually (17), the potential impact of heat wave on the incidence of diarrhoeal disease in the future could be concerning for Bangladesh. Given that the capital city of Dhaka is struggling to ensure water quality and facing a number of challenges to ensure the quality of urban life and sustainable urban growth including rising surface temperature in the context of urbanisation and global climate change, insufficient infrastructure, inadequate sanitation and poor hygiene brought about by poverty (18, 19), the impacts of heat waves on diarrhoea are likely to be considerably higher in Dhaka.

This paper aims to evaluate the influence of heat waves on hospitalisations due to diarrhoea in Dhaka. Although it is acknowledged that correlations uncovered do not necessarily imply direct causation, such indicators support understanding of the effects of heat waves on diarrhoeal disease morbidity thereby aiding further research to elicit linkages between climate change and gastrointestinal health. Given that many of the South Asian cities including Dhaka do not currently have a clearly agreed heat wave definition, this paper additionally aimed to identify pragmatic definitions of heat waves for Dhaka, which is a necessary first step to inform the development of a HWS for Dhaka.

Data and methodology

Diarrhoea data

Daily diarrhoea hospitalisation data between 1 January 1981 to 31 December 2010 were collected from the Dhaka Hospital of the International Centre for Diarrhoeal Diseases Research, Bangladesh (icddr,b) on 7 October 2020. The hospital served an urban population of approximately 3.5 million in 1981, 6.6 million in 1990 and 14.6 million in 2010 and provided free treatment to more than 140,000 patients with diarrhoea in 2010 (20). Given that reliable records of the total number of patients with admitted with diarrhoea per day or their disease onset dates were not available for the study period (1981–2010), information from the robust Diarrhoeal Disease Surveillance System (DDSS) was obtained instead to estimate the total number of patients hospitalised with diarrhoea per day. We did not access any information that could identify individual participants during or after data

collection. The DDSS platform recorded the information of all-cause diarrhoea patients who were enrolled into the surveillance system (21). It is likely that predominantly infectious gastroenteritis (IG) cases were included in this study. However, a limited number of people who had chronic or persistent diarrhoea at their first presentation and people with inflammatory bowel disease (IBD) who presented with similar symptoms were also likely included. Since ambient temperature including heat wave affect both IG and IBD (14), and because it was logistically impossible to test all stool samples for all possible pathogens, a syndromic approach was regarded appropriate for this study (S1: Additional information on health data).

Meteorological data

We collected data on daily climate parameters including the ambient, maximum, minimum temperature, cumulative rainfall, and relative humidity for Dhaka City from the Bangladesh Meteorological Department (BMD) from 1981 – 2010. The BMD recorded 3-hourly data from three validated weather stations for Dhaka (<https://bmd.gov.bd/external-link/https://dataportal.bmd.gov.bd/>).

Defining heat wave for Dhaka

In the absence of an acceptable and agreed definition of heat wave for Dhaka, Bangladesh, 16 indices of heat wave were calculated for Dhaka by incorporating the conditions known to affect thermal stress including day and night time temperature and duration and based on available data and resources (11). These are summarised in Table 1. Analyses of heat extremes were restricted to the

warm seasons (pre-monsoon summers and rainy monsoons – March – October) to avoid confounding by cold temperature (15).

Table 1: Definitions of the 16 proposed heat wave indicators tested. Max and min represent the daily maximum and minimum temperature, respectively. All indices calculated from March to October during 1981 – 2010

Index name	Conditions	Minimum duration (day)
TAV95	Daily mean temperature > 95 th percentile	1
TAV99	Daily mean temperature > 99 th percentile	1
D95	Daily max temperature > 95 th percentile	1
D99	Daily max temperature > 99 th percentile	1
MIN95	Daily min temperature >95 th percentile	1
D&N	Daily max and min temperature >95 th percentile	1
TAV952	Daily mean temperature > 95 th percentile	2
TAV953	Daily mean temperature > 95 th percentile	3
TAV992	Daily mean temperature > 99 th percentile	2
TAV993	Daily mean temperature > 99 th percentile	3
D952	Daily max temperature > 95 th percentile	2
D953	Daily max temperature > 95 th percentile	3
D992	Daily max temperature > 99 th percentile	2
D993	Daily max temperature > 99 th percentile	3
MIN952	Daily min temperature >95 th percentile	2
D&N2	Daily max and min temperature >95 th percentile	2

Exploratory analysis

Any missing data on the climate or health parameters were replaced by the by the respective month's average value for the parameter. Using established methods, each data series were checked for stationarity, autocorrelation, long-term trends, seasonality, possible outliers, normality, homoscedasticity and volatility (22-24).

Regression modelling

Negative binomial time series regression models were employed to compute the incidence rate ratio (IRR) estimates for the effect of heat waves on daily diarrhoea hospitalisations. The Wald-type 95% confidence intervals for the incidence rate ratios and associated *P*-values based on a reference distribution were also computed. The simple heat wave indicators defined above were used as predictor variables, and the regression models were used to determine the percentage increase or decrease in diarrhoea morbidity associated with each indicator. The risk estimates were adjusted for day of the week effects (with 7 categories, treating public holidays as Fridays), long-term time trend and seasonality (using natural cubic splines) and autocorrelation. Given that the optimum degree of freedom per year to account for the long-term trend and seasonality was unknown, the analysis was repeated with 3–7 degrees of freedom per year. The model with the lowest BIC value was the preferred model.

Past studies have shown significant effects of heavy rainfall and inconsistent effects of relative humidity on diarrhoea (25-34). As a result, the heat wave models were adjusted for heavy rainfall and humidity was not included in the models. Heavy rainfall (defined as the rainfall above the 95th percentile for the study period) was included as a categorical variable. Past studies have also highlighted potential lag effects of heavy rainfall on diarrhoea. Since individual and distributed lagged models allowed investigation of potential harvesting effects, correlation analysis was performed with relevant lag values of temperature extremes. As statistically significant relationship between heat wave at lags of 0 and 3 days and diarrhoea

were found, lag effects were considered in the final model. Lagged effects of heavy rainfall (0-8 days) were also included into the model. Ultimately, a constrained distributed lag linear model (DLLM) was used to investigate the effects of heat waves on diarrhoea after adjusting for the potential confounding effects of other meteorological factors, long-term trend, seasonality, day-of-the-week effect and autocorrelation.

The model took the following form:

$$Y_t \sim \text{Negative Binomial}(\mu_t, \theta)$$

$$\log[E(Y_t)] = \beta_0 + \sum \beta_{1p} ET_{t-1p} + \sum \beta_{2q} \text{HeavyRain}_{t-2q} + \sum NS(\text{time}_t, 7DF) + \sum \beta_3 \text{Dow} \quad (1)$$

Where, Y_t denotes daily all-cause diarrhoea count, ET_t and HeavyRain_t denote heat extreme and heavy rainfall indicator at time t . To control for long-term trends and seasonality, a natural cubic spline with 7 degrees of freedom per year was incorporated into the model. Dow_t was the categorical day of the week with a reference day of Friday.

The relative risk of hospitalisation for all-cause diarrhoea during a heat wave day was calculated from equation (1) as incidence rate ratio (IRR) and the associated percentage increase in hospitalisation during heat wave days were derived from the model parameters through Eq. (2)

$$\% \text{ change} = 100 (IRR - 1) \quad (2)$$

Multiple sensitivity analyses by changing the amount of control for seasonality and long-term trend, including relative humidity as a linear term and heavy rainfall as

a categorical variable without any lagged effects were carried out to check if the main findings were robust to changes in key assumptions. In addition, the analyses were rerun using the total number of diarrhoea patients enrolled into the icddr,b DDSS as the outcome instead of the total estimated diarrhoea hospitalisations per day.

Results

Between March to October 1981–2010, a total of 61,054 diarrhoea cases were enrolled into the DDSS platform and an estimated total of 2,171,500 patients of all ages and 1,103,325 children <5 years of age with all-cause diarrhoea sought hospital care from the icddr,b Dhaka Hospital. The average seasonal cycles of diarrhoea hospitalisation, temperature, rainfall and relative humidity in Bangladesh are shown in Fig 1. Diarrhoea hospitalisation in all ages and <5 children reached an annual maximum in April. Mean temperature remained high from April to June peaking during May before lowering down in October. Maximum temperature reached an annual maximum in April and May (close to 35°C) and decreased markedly during the rainy monsoon (June/July through to October) when the relative humidity was also high. However, night time temperatures (daily minimum) did not show similar pattern as day time temperatures (daily maximum). Relative humidity reached an annual maximum at approximately 90% during July and decreased towards the end of the rainy season in October.

Fig 1. Monthly distributions of all-cause diarrhoea hospitalisations in all ages (upper left) and <5 children (upper right), mean (upper middle left), maximum (upper middle right), minimum temperatures (lower middle left), relative humidity (lower middle right), cumulative rainfall (lower left) and heavy rainfall (lower right) in Dhaka, Bangladesh between 1 January 1981 and 31 December 2010

Fig 2 shows the temporal distribution of the heat wave day indicators by months. Most heat wave days were concentrated during the summer months. TAV95 and TAV99 heat wave categories peaked in May. D95 and D99 heat waves peaked during April. The MIN95 heat wave days were more widely distributed between April through October with the highest number found in June. The combined minimum and maximum temperature category (D&N) heat wave days were concentrated during April through June with the highest number in May.

Fig 2. Monthly distribution of heat wave days in Dhaka, Bangladesh, March–October 1981–2010

Table 2 displays the temporal distribution of the heat wave day indicators by decades. While the TAV95, TAV 99, MIN95 and D&N heat waves appeared to be increasing, the D95 and D99 heat wave days showed a decreasing trend across the decades. Table 3 shows the persistence of the heat wave day indicators lasting for 1–13 days. In all categories, most of the heat waves lasted for one day only with very few events lasting for more than four days. Two episodes of TAV95 heat wave and one episode of D95 heat wave lasting for a maximum of 13 consecutive days were identified during the study period.

Table 2. Distribution of the 16 proposed heat wave indicators tested by decades.

Max and min represent the daily maximum and minimum temperature, respectively. All indices calculated from March to October during 1981 - 2010

Index	Number of heat-wave events			
	1981-1990	1991-2000	2001-2010	1981-2010
TAV95	112	130	167	409
TAV99	20	26	28	74
D95	141	144	106	391
D99	45	34	11	90
MIN95	100	127	158	385
D&N	17	18	28	63
TAV952	24	29	31	84
TAV953	14	17	18	49
TAV992	6	5	6	17
TAV993	0	3	4	7
D952	28	30	23	81
D953	18	20	14	52
D992	10	6	1	17
D993	5	3	1	9
MIN952	17	23	31	71
D&N2	4	3	5	12

Table 3 Duration of persistence of heat wave days in Dhaka, Bangladesh, 1981–2010

Duration of persistence of heat wave (Days)	Heat wave category					
	TAV95	TAV99	D95	D99	MIN95	D&N
1	92	30	68	35	145	28
2	35	10	29	8	36	5
3	19	4	17	1	12	4
4	9	3	8	5	10	2
5	7	-	10	2	5	1
6	5	-	7	1	4	-
7	1	-	3	-	3	-
8	2	-	3	-	-	-
9	3	-	2	-	-	-
10	-	-	-	-	1	-
11	1	-	-	-	-	-
12	-	-	1	-	1	-
13	2	-	1	-	-	-

Table 4 displays the percentage increase in diarrhoea hospitalisation during heat wave events. We found significant increase in diarrhoea hospitalisation in all ages for only 5 out of the 16 proposed heat wave indicators. For <5 children significant results were obtained for 6 out of the 16 proposed indices. Compared to a non-heat wave day, all-cause diarrhoea hospitalisation increased by 7% and 8% in all ages and by 14% and 24% in children under 5 years on a TAV95 and TAV99 heat wave day, respectively. Increases in diarrhoea hospitalisations were strongest when defining heat waves using 99th percentile of daily maximum temperature.

Table 4. Percentage increase in diarrhoea hospitalisation in all ages and <5 children during heat wave days compared to non-heat wave days in Dhaka, 1981–2010

Indicator	All ages		<5 Children	
	Percentage increase in diarrhoea hospitalisations on heat wave days (95% CI)	P-value	Percentage increase in diarrhoea hospitalisations on heat wave days (95% CI)	P-value
TAV95	6.7 (4.6 – 8.9)	<0.001	13.9 (8.3 – 19.9)	<0.001
TAV99	8.3 (3.7 – 13.1)	<0.001	24.2 (11.3 – 38.7)	<0.001
D95	7.0 (4.8 – 9.3)	<0.001	17.0 (11.0 – 23.5)	<0.001
D99	7.4 (3.1 – 11.9)	0.001	19.5 (7.7 – 32.6)	0.001
MIN95	0.05 (-0.2 – 2.1)	0.964	4.4 (-0.8 – 9.9)	0.098
D&N	4.0 (-0.8 – 9.1)	0.107	14.0 (0.9 – 28.7)	0.035
TAV952	4.6 (0.4 – 9.0)	0.031	21.0 (3.2 – 41.9)	0.019
TAV953	-1.3 (-9.3 – 7.5)	0.770	17.0 (-5.7 – 45.2)	0.153
TAV992	1.9 (-0.50 – 9.2)	0.599	29.4 (-3.9 – 74.3)	0.089
TAV993	5.2 (-3.8 – 15.1)	0.269	16.7 (-27.0 – 86.5)	0.519
D952	1.9 (-5.0 – 9.2)	0.599	13.4 (-4.91 – 35.1)	0.162
D953	5.2 (-3.8 – 15.1)	0.269	22.3 (-2.7 – 53.6)	0.084
D992	11.5 (-2.3 – 27.2)	0.755	22.77 (-12.1 – 71.3)	0.230
D993	-34.1 (-5.2 – 36.5)	0.776	-11.8 (-65.4 – 25.2)	0.793
MIN952	-5.8 (-11.6 – 0.2)	0.058	-2.3 (-16.5 – 14.4)	0.755
D&N2	-4.8 (-19.4 – 12.5)	0.562	19.4 (-21.6 – 81.9)	0.409

*Bold values indicate significant results

Although lower than the same day effect, heat waves persisting for two days (TAV952) was significantly associated with diarrhoea among all ages. This effect was four times stronger in <5 children compared to all ages (4.6% Vs 21%). Significant effects were also observed for maximum temperature categories (D95 and D99). For all ages, neither the minimum temperature nor the days when both minimum and maximum temperature exceeded the 95th percentile (D&N) were found to be significantly associated with diarrhoea hospitalisation. However, significant effects of D&N were observed among <5 children. No significant effects of heat wave that lasted for three or more days were observed in these models.

Lagged effects of heat wave days were evaluated for 0–14 days initially in individual lag distributed models and later using constrained distributed lag linear models. Diarrhoea hospitalisation decreased by 3.5% (95% CI: 1.5% – 5.4%) three days following a TAV95 heat wave day. Significant negative effects of heavy rainfall were observed at lags 0-1 whereas significant positive effects of heavy rainfall were observed at the lags of 2–8 days. Compared to the holiday of week (Friday), diarrhoea hospitalisations were significantly higher in all weekdays with the highest effect observed on Sunday, when diarrhoea hospital increased by 10.3% (Table 5). Similarly, diarrhoea hospitalisation decreased by 4.9% (95% CI: 0.7% – 9.0%) three days following a TAV99 heat wave day.

Table 5 Adjusted associations among TAV95 heat wave (defined as the days with elevated mean temperature above the 95th percentile) and diarrhoea hospitalisations in Dhaka, March to October 1981 to 2010^a

Variable	IRR	95% CI	P-value
TAV95 Heat wave (Daily mean temperature >95th percentile)			
Lag 0	1.0672	1.0460 – 1.0889	0.000
Lag 1	1.0011	0.9765 – 1.0263	0.930
Lag 2	0.9783	0.9542 – 1.0030	0.084
Lag 3	0.9650	0.9457 – 0.9847	0.001
Heavy rainfall (>95th percentile)			
Lag 0	0.9098	0.8945 – 0.9253	0.000
Lag 1	0.9383	0.9221 – 0.9548	0.000
Lag 2	1.0201	1.0027 – 1.0378	0.023
Lag 3	1.0553	1.0373 – 1.0736	0.000
Lag 4	1.0544	1.0365 – 1.0727	0.000
Lag 5	1.0447	1.0270 – 1.0627	0.000
Lag 6	1.0243	1.0124 – 1.0478	0.001
Lag 7	1.0268	1.0076 – 1.0429	0.005
Lag 8	1.0222	1.0046 – 1.0394	0.013
Day of the week			
Friday	Referent		
Saturday	1.0759	1.0580 – 1.0941	0.000
Sunday	1.1033	1.0849 – 1.1221	0.000
Monday	1.0668	1.0491 – 1.0848	0.000
Tuesday	1.0368	1.0197 – 1.0543	0.000
Wednesday	1.0558	1.0383 – 1.0735	0.000
Thursday	1.0646	1.0470 – 1.0824	0.000

^a Constrained distributed lag linear model developed using equation 1 after controlling for long term trend and seasonality, autocorrelation, and lagged effects of heavy rainfall (0-8).
AIC=75800; BIC=75964; Dispersion statistic=0.9577; Mean deviance residual=-0.0573

Each model was evaluated to check model fit in addition to evaluating the dispersion statistic, and AIC and BIC values. Fig 3 displays the partial autocorrelation plot of deviance residuals from the final regression model depicting the relationship between TAV95 heat wave day and diarrhoea hospitalisation showing minimal residual autocorrelations.

Fig 3. Partial autocorrelation function plot of deviance residuals of the final regression model adjusted for autocorrelation where heat wave was defined by the exceedance of 95th percentile of the mean temperature

Discussion

This study found a statistically significant relationship between heat waves and diarrhoea hospitalisations in Dhaka, Bangladesh. This is one of the few studies to investigate the effects of heat waves on diarrhoeal disease morbidity and therefore provides essential information for analysing the potential impact of climate change on diarrhoea (14, 35). Diarrhoea hospitalisation increased by 6.7% (95% CI: 4.6% – 8.9%) and 8.3% (95% CI: 3.1% – 13.1%) on a TAV95 and TAV99 heat wave day.

On the other hand, diarrhoea hospitalisation decreased by 3.5% and 4.9% three days following a TAV95 and TAV99 heat wave day, respectively. The apparent protective incidence rate ratio obtained at the lag of 3 days suggested some degree of short-term morbidity displacement i.e. ‘harvesting’ effect. During heat waves, excess hospitalisation due to recent heat wave day (lag 3) may be offset by deficits due to diarrhoea hospitalisation accelerated a couple of days by previous heat wave days. A study in Vietnam has reported short-term displacement effect of diarrhoeal diseases due to rainfall (36). While a few previous studies have reported short-term displacement of deaths due to heat (37-39), the present study was the first to detect any harvesting effect due to extreme heat on diarrhoea hospitalisation. A previous

study investigating the effect of heat waves on infectious diarrhoea in Zurich reported a more pronounced effect of heat wave when a 7-day delayed effect of heat waves was considered. The same study reported an immediate effect of heat wave on diarrhoea due to inflammatory bowel disease (IBD). However, no harvesting effect of heat waves was identified in that study (31). In contrast, this study identified an both immediate and harvesting effect of heat waves on diarrhoea hospitalisation in Bangladesh.

Environmental temperature is known to play a vital role in the growth and replication of pathogenic bacteria including enterohaemorrhagic *Escherichia Coli* (EHEC) and influence bacterial composition on food, water and skin (14, 40, 41). Heat waves can promote environmental expansion of diarrhoeal pathogens, increase consumption of contaminated drinking water and/or increase food spoilage leading to excess diarrhoea. Given that a few previous studies have reported lags of days between dates of onset of diarrhoea and healthcare seeking in affected individuals, heat wave driven diarrhoea hospitalisation may be expected to take a few days to occur. However, the effects of extreme temperature on diarrhoea hospitalisation were mostly immediate in this study. One previous study using data from the same hospital in Dhaka reported that most of the severely dehydrated patients presented to the hospital within a narrow window of only 4–12 hours after symptom onset (42). This suggested that the hospitalised patients in this study likely presented to the hospital on the same day of the symptom onset. This may partly explain the observed immediate effect of heat waves on diarrhoea hospitalisation in Dhaka. In addition, heat waves may aggravate infectious diarrhoea among already affected individuals leading to the excess hospitalisation for diarrhoea on the same day. While most of the patients enrolled in this study are likely to be infectious in origin, a

few IBD diarrhoea cases may have been enrolled. Given that physical and mental stress can lead to flares of IBD and because heat stress are known to increase the frequencies of stress-dependent events including heart attacks and heat strokes (43), heat waves may trigger the flares of IBD or worsen a clinically non-apparent flare leading to excess diarrhoea (44).

In general, the effects of heat waves were most intense for children under 5 years of age compared to all ages. While the exact mechanism by which extreme temperature affect children's vulnerability to diarrhoea has never been investigated in much detail, children may be generally more susceptible to infections owing to their immature immune systems and low self-care capacity (45-48).

In this study, heat wave days defined by the exceedance of both 95th and 99th percentile of both daily mean temperature and daily maximum temperature performed as significant predictors of diarrhoea hospitalisation. A previous study investigating the effects of heat waves on mortality proposed the heat wave indicator combining day and night time temperatures as a suitable catchall indicator for heat waves in Bangladesh (15). However, D&N heat wave day was only significantly associated with childhood diarrhoea in the present study. The findings of the present study therefore suggest that D&N heat wave may not serve as a suitable indicator for heat wave in Dhaka, Bangladesh in relation to diarrhoeal disease morbidity. Although high nighttime temperatures (i.e. daily minimum temperature) are known to precipitate heat-related mortality by providing no cooling-down period at night (11), such effects may not be relevant to diarrhoeal disease context and expectedly high minimum temperature was not found to be significantly associated with diarrhoea

hospitalisation in this study. In addition, duration of heat waves was found to be less important when considering the effect of heat waves on diarrhoeal disease morbidity.

While the robust surveillance system, 30-year duration and the relative completeness of coverage of Dhaka's population constitute key strengths of the data set used in this study, there are several limitations. The estimated total number of all-cause diarrhoea cases hospitalised per day may not represent the exact number of cases admitted in the icddr,b Dhaka Hospital. Furthermore, the less severe cases would be less likely to be included. However, these issues do not pose a threat to the validity of the analysis of trends and comparisons over time, which is the theme of this study. In addition, numerous models were evaluated in this study during the sensitivity analysis to check the robustness of the results. Although the robustness of the results to varying degrees of control for long-term trend and seasonality was reassuring, yet there remains some possibility of residual confounding. Furthermore, there are uncertainties related to the extrapolation of the relationships revealed in this study to other locations with different climate and geography. In particular, the observed association may also be greatly dependent on the degree of water and sanitation infrastructure and hygiene practices in an area. Future studies from different geographic locations and socio-economic settings may provide additional information if the findings would pertain to other places.

Heat wave effects may vary depending on their intensity, duration, timing during the season and other traits as previous studies investigating heat wave effects on mortality suggested that heat waves occurring earlier in the summer can have higher effects on mortality. While the effects of intensity and duration of heat waves on diarrhoea has been examined in this study, the role of timing of heat

waves during the season was not investigated. While it is unlikely to change the results of this study significantly, future studies could benefit from defining heat waves using month-specific percentiles to examine the effect of timing of heat wave on diarrhoea.

Conclusion

This study identified heat wave as a risk factor for diarrhoea hospitalisation in Dhaka, Bangladesh by proposing several heat wave indices. TAV95 is the preferred heat wave indicator, which defines a heat wave as the elevated daily mean temperature above the 95th percentile persisting for at least one day. This definition results in 409 heat wave days and 176 separate heat waves in 30 years from 1981 to 2010. Almost all the heat waves occurred during the pre-monsoon summer season, between April and June, with the highest number of heat waves in May. Diarrhoea hospitalisations increased by 7% in all ages and 14% among children under 5 years of age during a TAV95 heat wave day compared to a non-heat wave day. These results can be used to define heat waves for Dhaka and motivate public health interventions including generation of heat alerts to prevent heat-related morbidity in Dhaka.

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557 **Supporting information**

558 S1: Additional information on health data

559

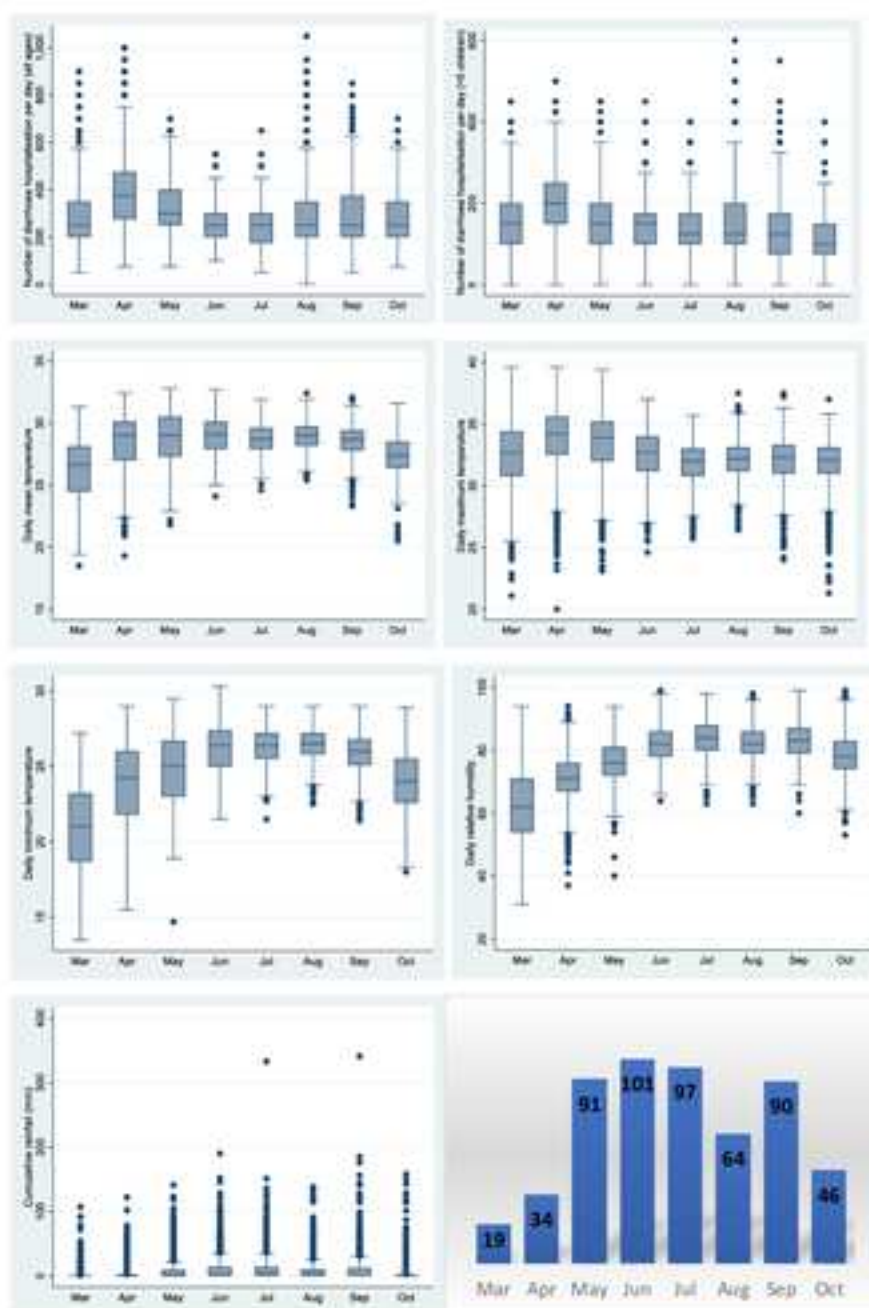


Fig. 1. Monthly distributions of all-cause diarrhoea hospitalisations in all ages (upper left) and <5 children (upper right), mean (upper middle left), maximum (upper middle right), minimum temperatures (lower middle left), relative humidity (lower middle right), cumulative rainfall (lower left) and heavy rainfall (lower right) in Dhaka, Bangladesh between 1 January 1981 and 31 December 2010

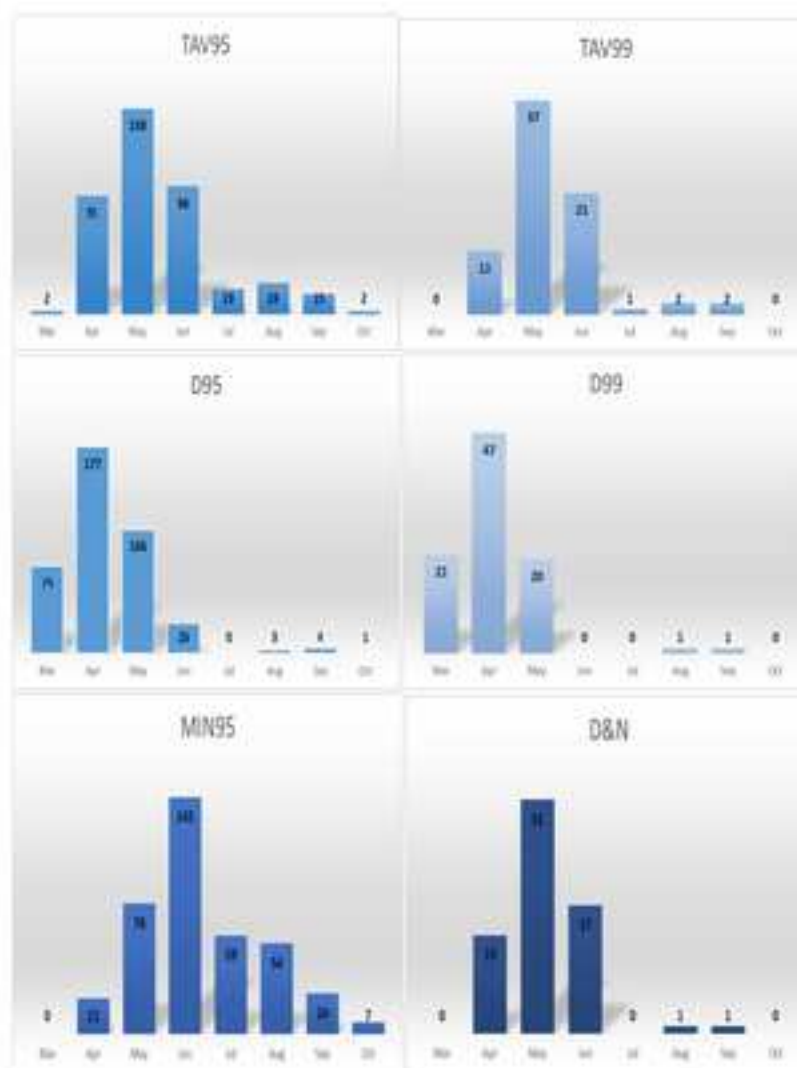


Fig. 2. Monthly distribution of heat wave days in Dhaka, Bangladesh, March–October 1981–2010

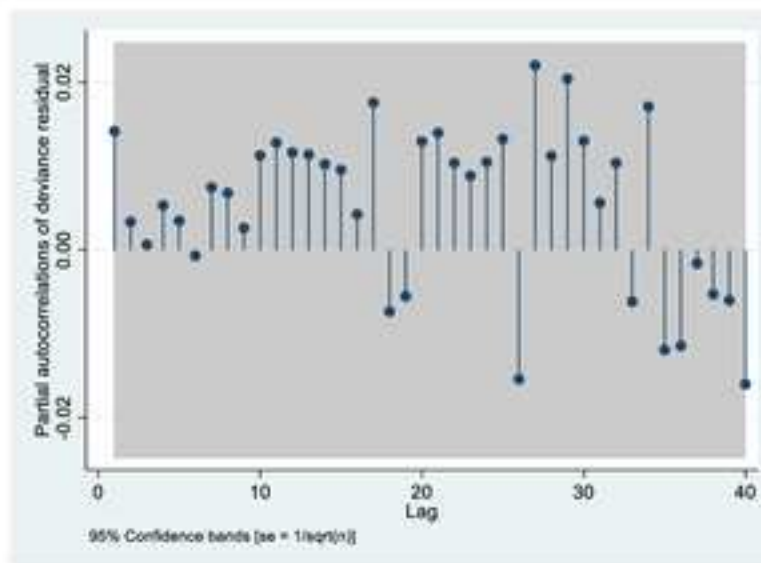
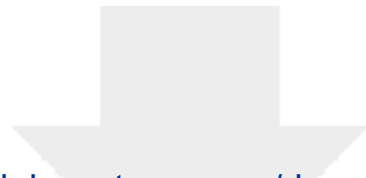


Fig. 3. Partial autocorrelation function plot of deviance residuals of the final regression model adjusted for autocorrelation where heat wave was defined by the exceedance of 95th percentile of the mean temperature



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