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Gideon O. Emukule,a,b Joshua A. Mott,a,c,d Peter Spreeuwenberg,e Cecile Viboud,f Alexander Commanday,c Philip Muthoka,g Patrick K. Munywoki,h David J. Nokes,h,i Koos van der Velden,b John W. Pagetb,e

aCenters for Disease Control and Prevention – Kenya Country Office, Nairobi, Kenya. bDepartment of Primary and Community Care, Radboud University Medical Center, Nijmegen, The Netherlands. cInfluenza Division, National Center for Immunization and Respiratory Diseases, US Centers for Disease Control and Prevention, Atlanta, GA, USA. dUS Public Health Service, Rockville, MD, USA. eNetherlands Institute for Health Services research (NIVEL), Utrecht, The Netherlands. fFogarty International Center, National Institutes of Health, Bethesda, MD, USA. gMinistry of Health, Government of Kenya, Nairobi, Kenya. hKenya Medical Research Institute, Centre for Geographic Medicine Research-Coast, Kilifi, Kenya. iSchool of Life Sciences, University of Warwick, Coventry, UK.

Correspondence: Gideon O. Emukule, KEMRI Headquarters, Mbagathi Road, Off Mbagathi Way, PO Box 606, Village Market 00621, Nairobi, Kenya. E-mail: uyr9@cdc.gov

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Background Information on the timing of influenza circulation remains scarce in Tropical regions of Africa.

Objectives We assessed the relationship between influenza activity and several meteorological factors (temperature, specific humidity, precipitation) and characterized the timing of influenza circulation and its implications to vaccination strategies in Kenya.

Methods We analyzed virologically confirmed influenza data for outpatient influenza-like illness (ILI), hospitalized for severe acute respiratory infections (SARI), and cases of severe pneumonia over the period 2007–2013. Using logistic and negative binomial regression methods, we assessed the independent association between climatic variables (lagged up to 4 weeks) and influenza activity.

Results There were multiple influenza epidemics occurring each year and lasting a median duration of 2–4 months. On average, there were two epidemics occurring each year in most of the regions in Kenya, with the first epidemic occurring between the months of February and March and the second one between July and November. Specific humidity was independently and negatively associated with influenza activity. Combinations of low temperature (<18°C) and low specific humidity (<11 g/kg) were significantly associated with increased influenza activity.

Conclusions Our study broadens understanding of the relationships between seasonal influenza activity and meteorological factors in the Kenyan context. While rainfall is frequently thought to be associated with influenza circulation in the tropics, the present findings suggest low humidity is more important in Kenya. If annual vaccination were a component of a vaccination strategy in Kenya, the months of April to June are proposed as optimal for associated campaigns.

Keywords Humidity, influenza, Kenya, respiratory, seasonality, vaccination.

Introduction Influenza exerts a significant health burden on human populations across temperate, subtropical, and tropical regions.1,2 In temperate regions, influenza epidemics exhibit clear seasonality with peaks during winter months3,4 suggestive of an association with climatic factors. In these regions, lower temperature and lower specific humidity have been shown to be significantly associated with increased influenza activity.5,6 In contrast, influenza seasonal characteristics are less predictable in tropical and subtropical regions which are characterized by semiannual epidemics or year-round influenza activity.5,7–10 A meteorological factor that is frequently reported to be associated with high influenza incidences in the tropical areas is rainfall.8,9,11

In temperate countries, a well-defined seasonality allows for a precise timing of influenza vaccination campaigns to precede periods of peak circulation. However, in tropical African countries, more data are needed on influenza seasonality and its determinants. In Kenya, where there is
currently no influenza vaccination strategy in place, these data may help to inform vaccine implementation strategy decisions. Kenya experiences long rains that occur from March to May and short rains occurring in October and November. Temperatures are highest during the months of January to March. However, there is considerable climate variability within Kenya such that influenza surveillance has been set up in different locations, including the coastal region with average monthly temperatures ranging between 25 and 26°C during the hot season (July–August). The northern/northeastern regions have semiarid and desert-like conditions and experience lowest temperatures in the country (as low as 7°C during the July–August cold season). The central region has a relatively higher altitude compared to the other regions and experiences some of the lowest temperatures in the country (as low as 7°C during the July–August cold season). The northern/northeastern regions have semiarid and desert-like conditions and experience sunny and dry wintermost of the year. The coastal region experiences hot and humid weather conditions year round with average monthly temperatures ranging between 25 and 29°C and an average annual rainfall of over 1000 mm. The environmental data used in this analysis were satellite-derived measurements and were collected over the same period of time as the influenza data. These variables were calculated the proportion of specimens that tested positive for influenza A and B viruses. Influenza A-positive specimens were subtyped for A(H1N1), A(H3N2), and A(H1N1) pdm09.

### Influenza activity patterns

To assess influenza circulation over time, we calculated the average monthly proportion of influenza-positive cases among those tested across the study period. Data from the pandemic period (August 2009–July 2010) were excluded because (i) these were likely to influence the pattern of results and (ii) we were interested in seasonal influenza. We also calculated the proportion of specimens that tested positive for influenza in the first half of the year within which the first epidemic typically occurs (January–June), and in the second half of the year within which the second epidemic occurs (July–December). Data from all the 12 surveillance sites were included in this analysis.

### Defining “influenza circulation periods”

“Influenza circulation periods” were defined as a period of ≥2 successive weeks where ≥10% of the total weekly cases tested were positive for influenza. The 10% threshold was close to the average proportion (12%) of cases that tested positive for influenza over the entire study period. In situations where there were <25 cases tested in a week, we considered the proportions as unstable and used the five-point moving average method to estimate the number of influenza cases and the percentage positive for influenza. Influenza circulation periods separated by ≥5 successive weeks where there was low influenza activity (<10%) were considered as two distinct periods. The first week of influenza circulation period is herein referred also as the start-week or onset week (Figure S1).

### Meteorological data

The environmental data used in this analysis were satellite-derived measurements and were collected over the same period of time as the influenza data. These variables were average surface temperature (°C) and near-surface specific humidity (g/kg) obtained from the Global Land Data Assimilation System (GLDAS), and accumulated rainfall (mm) obtained from the Tropical Rainfall Measuring Mission (TRMM) (Appendix S1).

### Data analyses

Descriptive analyses for influenza and meteorological factors

Influenza circulation was described using proportions of influenza A- and/or B-positive cases. The age distribution...
and the influenza activity patterns were described using medians and ranges. The meteorological factors were described using means and standard deviations (SD).

Bivariate and multivariate analyses of influenza activity

Data from nine of the 12 sites were used when assessing the associations between influenza activity and meteorological variables. Three surveillances sites (Kenyatta National Hospital, Mombasa County Referral Hospital, and Ting’wang’i Health Center) were excluded from these analyses because of multiple missing data points in the time series (Appendix S1).

We applied two analytical approaches to determine the association between influenza activity and meteorological variables: (i) logistic regression to determine the association between the onset of influenza activity and meteorological variables and (ii) negative binomial regression to determine the association between the weekly number of influenza cases and meteorological variables. In the first approach, the binary outcome variable “start-week” was coded as “1” if the week considered was the onset week of influenza activity or otherwise coded as “0.” In the second approach, the outcome variable was the weekly count of influenza-positive cases identified at each site. Negative binomial regression was chosen over the over Poisson regression to account for overdispersion in the data. The variables that were considered as covariates in the models were site, year, and week of the year (week = 1, 2, 3, . . ., 52). With the exception of the site and year variables which were analyzed as categorical variables, all the other variables were entered into the respective models as continuous variables.

We investigated associations of up to 4 lagged weeks on all meteorological variables to assess a possible delayed weather effect on influenza activity (Appendix S1). We additionally assessed whether “cold–dry” and “humid–rainy” conditions – as suggested in a recent global seasonality
Table 1. Descriptive statistics for influenza testing, January 2007-December 2013

<table>
<thead>
<tr>
<th>Hospital/Clinic</th>
<th>Period of data included in analysis</th>
<th>Median age (yr)</th>
<th>Number tested positive</th>
<th>Number hospitalized positive</th>
<th>SARI cases tested positive</th>
<th>SARIAI cases tested positive</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lwak Hospital (Kenya)</td>
<td>2007-2013</td>
<td>50.0 (±24.2)</td>
<td>1162 (10.7)</td>
<td>273 (23.6)</td>
<td>485 (41.5)</td>
<td>355 (30.4)</td>
</tr>
<tr>
<td>Eastern Kenyan Health Centre</td>
<td>2008-2013</td>
<td>57.0 (±21.3)</td>
<td>876 (9.6)</td>
<td>228 (25.8)</td>
<td>440 (50.1)</td>
<td>325 (37.2)</td>
</tr>
<tr>
<td>Lwak hospital (Kilifi)</td>
<td>2008-2013</td>
<td>48.0 (±25.0)</td>
<td>263 (2.1)</td>
<td>64 (24.2)</td>
<td>132 (50.0)</td>
<td>105 (40.2)</td>
</tr>
<tr>
<td>Kilifi Hospital (Kilifi)</td>
<td>2008-2013</td>
<td>63.0 (±20.2)</td>
<td>752 (9.4)</td>
<td>192 (25.5)</td>
<td>376 (50.0)</td>
<td>282 (37.3)</td>
</tr>
<tr>
<td>Kilifi Hospital (Moira)</td>
<td>2008-2013</td>
<td>57.0 (±20.0)</td>
<td>151 (1.7)</td>
<td>37 (24.6)</td>
<td>78 (51.7)</td>
<td>61 (40.3)</td>
</tr>
<tr>
<td>Kilifi hospital (Kilifi)</td>
<td>2008-2013</td>
<td>57.0 (±20.0)</td>
<td>275 (3.2)</td>
<td>70 (25.6)</td>
<td>145 (52.7)</td>
<td>113 (41.3)</td>
</tr>
<tr>
<td>Mombasa Hospital (Mombasa)</td>
<td>2007-2013</td>
<td>60.0 (±25.1)</td>
<td>480 (3.3)</td>
<td>120 (25.0)</td>
<td>240 (50.0)</td>
<td>180 (37.5)</td>
</tr>
<tr>
<td>Mombasa Hospital (Mombasa)</td>
<td>2007-2013</td>
<td>60.0 (±25.1)</td>
<td>680 (3.3)</td>
<td>170 (25.0)</td>
<td>340 (50.0)</td>
<td>270 (37.5)</td>
</tr>
<tr>
<td>Mombasa Hospital (Mombasa)</td>
<td>2007-2013</td>
<td>60.0 (±25.1)</td>
<td>800 (3.3)</td>
<td>200 (25.0)</td>
<td>400 (50.0)</td>
<td>300 (37.5)</td>
</tr>
</tbody>
</table>

NA: not applicable.

*Acute lower respiratory illness.

**Population-based disease surveillance sites.

***At Kilifi CH, samples were collected from children 5 years who were hospitalized with severe or very severe pneumonia.
study—were associated with influenza activity.5 We used combinations of temperature and specific humidity at thresholds of <18°C and <11 g/kg, respectively, to define “cold–dry” conditions and combinations of specific humidity and rainfall at thresholds of >14 g/kg and >150 mm, respectively, to define “humid–rainy” conditions. In addition to investigating the effect of the “cold–dry” and “humid–rainy” conditions on influenza activity, we also assessed the effect of the two-way product interaction (included as continuous variables) between temperature and specific humidity, and between specific humidity and rainfall. The interactions were evaluated in the model alongside the main effects of temperature, specific humidity, and rainfall.

The multiple variable models for the logistic and negative binomial regression analyses were fitted by including the site variable as well as all the variables that were associated with influenza in the bivariate analysis at overall \( P \)-value < 0.2. Statistical significance was considered if the \( P \)-value was <0.05.

All data analyses were performed using Stata version 13.0 (StataCorp. 2013, Stata Statistical Software: Release 13. College Station, TX: StataCorp LP).

Ethical considerations
The study protocols were approved by both the institutional review board (IRB) of the U.S. CDC (CDC-3308, CDC-4566) and the ethical review committee of the Kenya Medical Research Institute (KEMRI; SSC-1801, SSC-932, SSC-1161, SSC-1055, 1526, 1858). At Nakuru, Kakamega, and Nyeri County Referral Hospitals, the Kenya Ministry of Health (KMoH) issued a letter stating that sentinel surveillance for influenza in the bivariate analysis at overall \( P \)-value < 0.2. Verbal consent at these sites was obtained from all patients before questionnaires were administered and specimens were collected. For children, verbal consent was obtained from guardians.

Results

Descriptive analyses
A total of 55 192 patients were tested for influenza at the 12 surveillance sites over the period 2007 to 2013 of which 6721 (12%) tested positive for influenza. The proportion of patients who tested positive for influenza ranged from 4% in Kilifi to 19% among patients who were seen at Dadaab refugee camp. The median age of the patients who were tested for influenza was 17 years [interquartile range (IQR) = 0–8–4·2 years; Table 1].

The mean average weekly temperature was lowest in Nakuru (18·2°C) and highest at Dadaab refugee camp (30·7°C; Table 2). The mean average weekly specific humidity ranged from 11·1 to 15·1 g/kg with Nyeri recording the lowest measurements, while Kilifi recorded the highest. The “cold–dry” conditions defined earlier were observed in only at two sites in the central Kenya region (Nakuru and Kibera). However, the “humid–rainy” conditions were only experienced at the coastal site (KCH) and at only four different time points (weeks) over the course of the study period.

Influenza activity patterns
A total of 48 periods of increased influenza circulation were identified across the nine study sites. Nineteen of these episodes occurred within the first quarter of the year (median onset month was February), 16 episodes occurred in the second quarter (median onset month was July), and the remaining 13 occurred in the last quarter (median onset month was October). On average, most of the study sites experienced two episodes of increased influenza circulation annually which lasted for a median duration of 2–4 months.

When we analyzed the monthly seasonal cycle of influenza activity, there was a pattern showing periods of increased influenza circulation occurring between February and March, and between July and November. There were more influenza-positive cases identified in the last half of the years included in the analysis—within which the second epidemic occurs—compared to the first half of the years [3886 (58%) influenza cases during July–December versus 2835 (42%) during January–June]. The month of May had the lowest influenza activity (Figure 2 and Figure S1).

Bivariate and multivariable analyses
In the bivariate models for onset of influenza activity, specific humidity was significantly and negatively associated with the onset of influenza activity \((P < 0.05)\). In the negative binomial regression models, influenza activity was found to be negatively associated with both temperature and specific humidity \((P < 0.05)\). The presence of the “cold–dry” conditions, defined earlier, was also found to be significantly associated with influenza activity \((P < 0.05)\). No statistically significant associations were observed between influenza activity and rainfall (Table 3).

In the multivariable logistic regression model, specific humidity was independently and negatively associated with the onset of influenza activity at lag weeks one \([\text{odds ratio (OR)} = 0.79 (95\% \text{CI} 0.66–0.94)]\) and two \([\text{OR} = 0.82 (95\% \text{CI} 0.69–0.98)]\) in the models that adjusted for the site variable. Similarly, specific humidity was significantly associated with influenza activity in the negative binomial regression models for the weekly count of influenza cases at the current week \([\text{incidence rate ratio (IRR)} = 0.94 (95\% \text{CI} 0.90–0.98)]\), and at all the four lag weeks investigated \((P < 0.001)\). The presence of “cold–dry” conditions was also found to be positively associated with influenza activity when we adjusted for the site variable at current week \([\text{IRR} = 1.90 (95\% \text{CI} 1.20–3.01)]\), and at lag weeks one \([\text{IRR} = 2.07 (95\% \text{CI} 1.20–3.01)]\).
<table>
<thead>
<tr>
<th>Hospital/clinic</th>
<th>Period of data included in analysis</th>
<th>Yearly number of influenza circulation periods</th>
<th>Temperature (°C)</th>
<th>Specific humidity (g/kg)</th>
<th>Accumulated rainfall (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td>Mean (SD)</td>
<td>Median (IQR)</td>
<td>Mean (SD)</td>
</tr>
<tr>
<td>St. Elizabeth Hospital (Lwak)</td>
<td>2007-2013</td>
<td>2</td>
<td>21.4 (1.5)</td>
<td>21.2 (20.3–22.3)</td>
<td>13.0 (1.3)</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Siaya County Referral Hospital (CRH)</td>
<td>2010-2013</td>
<td>2</td>
<td>21.2 (1.4)</td>
<td>20.9 (20.3–21.9)</td>
<td>13.0 (1.2)</td>
</tr>
<tr>
<td>Kakamega CRH</td>
<td>2008-2011</td>
<td>1</td>
<td>21.4 (1.6)</td>
<td>21.2 (20.3–22.5)</td>
<td>13.2 (1.3)</td>
</tr>
<tr>
<td>Ting’Wang’i Health Center*</td>
<td>N/A</td>
<td>–</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
</tr>
<tr>
<td>Western Kenya region</td>
<td>2007-2013</td>
<td>2</td>
<td>21.4 (1.5)</td>
<td>21.2 (20.3–22.3)</td>
<td>13.0 (1.2)</td>
</tr>
<tr>
<td>Kenyatta National Hospital*</td>
<td>N/A</td>
<td>–</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
</tr>
<tr>
<td>Tabitha Clinic (Kibera)</td>
<td>2009-2013</td>
<td>2</td>
<td>19.8 (1.3)</td>
<td>19.7 (18.8–20.7)</td>
<td>11.5 (1.3)</td>
</tr>
<tr>
<td>Nyeri CRH</td>
<td>2009-2012</td>
<td>2</td>
<td>19.8 (1.1)</td>
<td>19.6 (19.0–20.6)</td>
<td>11.1 (1.4)</td>
</tr>
<tr>
<td>Nakuru CRH</td>
<td>2009-2013</td>
<td>2</td>
<td>18.2 (1.2)</td>
<td>17.9 (17.3–18.8)</td>
<td>11.1 (1.5)</td>
</tr>
<tr>
<td>Central Kenya region</td>
<td>2009-2013</td>
<td>2</td>
<td>19.2 (1.2)</td>
<td>19.0 (18.4–20.0)</td>
<td>11.3 (1.3)</td>
</tr>
<tr>
<td>Dadaab refugee camp</td>
<td>2008-2009</td>
<td>2</td>
<td>30.7 (1.8)</td>
<td>30.9 (29.4–32.1)</td>
<td>13.4 (1.6)</td>
</tr>
<tr>
<td>Kakuma refugee camp</td>
<td>2007-2012</td>
<td>2</td>
<td>30.1 (1.7)</td>
<td>30.1 (29.1–31.1)</td>
<td>11.7 (2.1)</td>
</tr>
<tr>
<td>Northern/Northeastern region</td>
<td>2007-2012</td>
<td>2</td>
<td>30.0 (1.8)</td>
<td>30.1 (29.1–31.2)</td>
<td>12.1 (2.0)</td>
</tr>
<tr>
<td>Mombasa CRH*</td>
<td>N/A</td>
<td>–</td>
<td>N/A</td>
<td>N/A</td>
<td>N/A</td>
</tr>
<tr>
<td>Kilifi CH</td>
<td>2007-2013</td>
<td>1</td>
<td>27.6 (1.5)</td>
<td>27.5 (26.4–28.7)</td>
<td>15.1 (1.5)</td>
</tr>
<tr>
<td>Coastal Kenya region</td>
<td>2007-2013</td>
<td>1</td>
<td>27.6 (1.5)</td>
<td>27.5 (26.4–28.7)</td>
<td>15.1 (1.5)</td>
</tr>
<tr>
<td>All sites</td>
<td>2007-2013</td>
<td>2</td>
<td>24.77 (4.5)</td>
<td>25.5 (20.5–28.9)</td>
<td>13.0 (2.1)</td>
</tr>
</tbody>
</table>

N/A, not applicable.

*Data from these sites were excluded from the analysis of association of influenza activity and meteorological variables because of multiple missing data points in the time series.
CI 1.21–3.55) and three [IRR = 1.95 (95% CI 1.11–3.44)]. However, temperature was not significantly associated with influenza activity when we adjusted for the site variable. All the other variables assessed including rainfall and the two-way interactions between specific humidity and temperature and between specific humidity and rainfall were not significantly associated with influenza activity when we adjusted for the site variable (Table 4). An exploratory analysis to assess the relationship between the onset week of influenza activity and meteorological variables showed similar results to the ≥10% activity threshold when we used the median proportion (7% threshold) to define the onset of influenza activity (results not shown).

**Discussion**

In this study, we found that there were multiple periods of increased influenza activity annually in Kenya. On average, there were two epidemics occurring each year in most of the regions in Kenya and these epidemics lasted a median duration of 2–4 months. The first epidemic occurred between February and March and the second between July and November. The period between April and May had the least influenza activity. We also identified that lower specific humidity was significantly associated with influenza activity in Kenya. As has been noted in other continents, we found that influenza was more likely to circulate when both temperature and specific humidity were below 18°C and 11 g/kg, respectively, independent of the study site. Contrary to what has been hypothesized previously, we found no significant association between influenza activity and rainfall.

Unlike temperate climates, the presence of multiple influenza epidemics each year in most of the regions in Kenya presents a challenge to the selection of the appropriate influenza vaccine formulation to use. Recent investigations have found that the Southern Hemisphere (SH) vaccine formulation was well matched [80% (95% CI 77–84) over a 9-month period] to circulating strains over the period...
2007–2013 (Waiboci et al., accepted for publication in Vaccine). The Northern Hemisphere (NH) vaccine formulation was also well matched [82% (95% CI 78–85) over a 9-month period]. These findings suggest that for the primary period of increased influenza circulation in Kenya (July–November), the SH vaccine formulation (available in April)
could offer good protection. While this vaccine could also provide protection during the subsequent February and March peaks as well, the NH formulation (available in November) could also be considered for that period.

Our finding of a negative association between influenza activity and specific humidity is consistent with findings

<table>
<thead>
<tr>
<th>Year</th>
<th>Odds Ratio (95% CI)</th>
<th>P-value</th>
<th>Incidence Rate Ratio (95% CI)</th>
<th>P-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>2007</td>
<td>0.70 (0.23–2.15)</td>
<td>0.233*</td>
<td>0.45 (0.37–0.55)</td>
<td>&lt;0.001*</td>
</tr>
<tr>
<td>2008</td>
<td>0.74 (0.30–1.84)</td>
<td>0.516</td>
<td>0.78 (0.64–0.94)</td>
<td>0.010</td>
</tr>
<tr>
<td>2009</td>
<td>0.34 (0.10–1.18)</td>
<td>0.089</td>
<td>0.79 (0.62–1.00)</td>
<td>0.048</td>
</tr>
<tr>
<td>2010</td>
<td>0.31 (0.07–1.36)</td>
<td>0.120</td>
<td>0.96 (0.79–1.16)</td>
<td>0.649</td>
</tr>
<tr>
<td>2011</td>
<td>Ref</td>
<td></td>
<td>Ref</td>
<td></td>
</tr>
<tr>
<td>2012</td>
<td>0.99 (0.46–2.11)</td>
<td>0.979</td>
<td>0.60 (0.49–0.74)</td>
<td>&lt;0.001*</td>
</tr>
<tr>
<td>2013</td>
<td>0.42 (0.14–1.27)</td>
<td>0.125</td>
<td>0.69 (0.56–0.85)</td>
<td>&lt;0.001*</td>
</tr>
<tr>
<td>Week</td>
<td>0.98 (0.97–1.00)</td>
<td>0.123</td>
<td>1.00 (0.99–1.00)</td>
<td>0.285</td>
</tr>
<tr>
<td>Site</td>
<td>0.813*</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

St. Elizabeth Hospital
- Tabitha Clinic (Kibera)
  1.52 (0.56–4.12)
- Nyeri CRH
  0.75 (0.19–2.85)
- Kakamega CRH
  1.00 (0.30–3.37)
- Nakuru CRH
  1.13 (0.39–3.30)
- Siaya CRH
  1.14 (0.45–4.05)
- Dadaab refugee camp
  1.05 (0.38–2.94)
- Kakuma refugee camp
  0.924 (0.29–3.21)
- Kilifi County Referral Hospital
  0.49 (0.15–1.66)

Temperature (°C)
- No lag
  0.99 (0.93–1.05)
- Lag 1 week
  0.98 (0.92–1.04)
- Lag 2 weeks
  0.97 (0.91–1.04)
- Lag 3 weeks
  0.97 (0.91–1.04)
- Lag 4 weeks
  0.97 (0.91–1.04)

Specific humidity (g/kg)
- No lag
  0.86 (0.75–0.98)
- Lag 1 week
  0.82 (0.71–0.94)
- Lag 2 weeks
  0.84 (0.73–0.96)
- Lag 3 weeks
  0.89 (0.78–1.02)
- Lag 4 weeks
  0.86 (0.75–0.99)

Accumulated rainfall (mm)
- No lag
  1.00 (0.99–1.01)
- Lag 1 week
  0.99 (0.98–1.00)
- Lag 2 weeks
  1.00 (1.00–1.01)
- Lag 3 weeks
  1.00 (0.99–1.01)
- Lag 4 weeks
  1.00 (0.99–1.01)

Presence of cold-dry conditions**
- No lag
  8.17e-06 (0–)
- Lag 1 week
  8.17e-06 (0–)
- Lag 2 weeks
  8.17e-06 (0–)
- Lag 3 weeks
  8.17e-06 (0–)
- Lag 4 weeks
  8.18e-06 (0–)

*Overall P-value.
**Cold-dry periods were defined as weeks when the average temperature was <18°C and specific humidity was <11 g/kg.
from other studies that were conducted in temperate and subtropical regions. This is also consistent with experimental results which have linked low humidity to prolonged influenza virus survival (IVS) as well as efficient aerosol transmission. These findings also suggest a site-specific association between temperature and influenza activity as temperature was negatively and significantly associated with influenza activity in the bivariate analysis but not when we adjusted for the site variable. Whereas rainfall has previously been suggested to be correlated with influenza activity in tropical and subtropical regions, our study did not find a significant association. A recent global study found that “humid–rainy” (high specific humidity and rainy) conditions were associated with influenza circulation. However, our analysis did not support this finding. The lack of association between the “humid–rainy” conditions with influenza activity in our study context may in part be explained by the fact that we do not experience necessary thresholds of high specific humidity (>14 g/kg) and high rainfall (>150 mm) measurements as previously suggested. Indeed, these conditions were only experienced at four different time points over the course of the study period at the coastal site (KCH).

The relative merits of annual influenza vaccination versus the integration of influenza vaccination into routine immunization schedules remain to be evaluated in Kenya and are beyond the scope of this discussion. However, if annual mass vaccination campaigns are being considered, the period between April and June would perhaps be the optimal time for several reasons. First, this would potentially offer better protection considering the fact that the period of influenza activity between July and December account for most of the annual influenza cases (58%). Considering the possibility of waning immunity over time, it would probably be preferable to vaccinate during the month of June. However, a wider period may need to be considered in the context of the possible logistical challenges of vaccine delivery and accessing the target populations. Second, according to the Kenyan education calendar, schools are closed for holidays during the months of April, August, and December. April would therefore be a more convenient time for school-going children to be immunized in non-school settings. Lastly, caretakers of young children may take advantage of the possibility of vaccination campaigns being considered, the period

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<td><strong>Association with the onset of influenza activity</strong></td>
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<td><strong>Absolute association with influenza activity</strong></td>
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<td><strong>Temperature (°C)</strong></td>
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<td><strong>Specific humidity (g/kg)</strong></td>
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<td>No lag</td>
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<td>Lag 4 weeks</td>
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<td><strong>Accumulated rainfall (mm)</strong></td>
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<td>Lag 4 weeks</td>
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<td><strong>Presence of cold–dry conditions</strong></td>
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*Cold–dry periods were defined as weeks when the average temperature was <18°C and specific humidity was <11 g/kg. NE, not estimated.
matters such as taking the smaller children for immunization.

Our study was subject to some other important limitations. First, we were not able to account for the effect of other factors such as socioeconomic conditions, population susceptibility, and human migration dynamics on the association between influenza activity and meteorological variables because these data were not collected. Second, we only relied on satellite-derived meteorological measurements for our analysis. Even though we had a reasonable temporal resolution in the meteorological data, using actual ground data could possibly have provided more accurate results. Third, although we tried to adjust for the effect of the site differences in our models, we could not sufficiently explore the regional variation of meteorological factors in Kenya and how they correlate with influenza activity because of limited influenza testing data available. Lastly, we could not explore whether the association between influenza activity and meteorological factors varied by age as older persons were underrepresented in the hospital-based surveillance because of low healthcare seeking behavior.25

In conclusion, our study broadens our understanding of the relationships between seasonal influenza activity and meteorological factors in tropical regions, and more specifically in the Kenyan context. We additionally highlight the influenza activity patterns in Kenya with regard to the onset months of periods of increased influenza circulation. These could help to inform the timing of future influenza vaccination campaigns in Kenya and highlight periods when added diagnostic measures, treatment efforts, or infection control strategies may be put in place.

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Author contributions

Gideon O. Emukule, Joshua A. Mott, Peter Spreeuwenberg, Cecile Viboud, D. James Nokes, Koos van der Velden, and John W. Paget involved in concept and design of manuscript; Gideon O. Emukule analyzed the data and lead author in writing the manuscript; Gideon O. Emukule, Joshua A. Mott, Peter Spreeuwenberg, Cecile Viboud, Philip Muthoka, Patrick K. Munywoki, D. James Nokes, Koos van der Velden, and John W. Paget involved in interpretation of data; Gideon O. Emukule, Joshua A. Mott, Peter Spreeuwenberg, Cecile Viboud, Alexander Commanday, Patrick K. Munywoki, D. James Nokes, Koos van der Velden, and John W. Paget helped in writing of the manuscript; and Alexander Commanday involved in literature review.

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Disclaimer

The findings and conclusions in this report are those of the authors and do not necessarily represent the official position of the Centers for Disease Control and Prevention.

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Supporting Information

Additional Supporting Information may be found online in the supporting information tab for this article:

Figure S1. Figures showing the time series of the average weekly temperature, specific humidity and percent influenza positive cases by site, 2007 to 2013.
Appendix S1. Supplemental methods.