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26	Understanding the environmental drivers of influenza transmissibility would contribute
27	to the early intervention and long-term control strategies of seasonal influenza, a
28	serious public health problem that causes considerable morbidity and mortality each
29	year. Within the burgeoning literature on influenza transmission, there are conflicting
30	lines of evidence on the role of the environment [1]. Besides meteorological factors, it is
31	also uncertain how common air pollutants such as ozone (O_3), sulphur dioxides (SO ₂),
32	nitrogen dioxide (NO $_2$), nitric oxide (NO), and particulate matter (PM) may affect
33	influenza transmission [2]. The objective of our study was to examine the relationship
34	of influenza transmissibility in Hong Kong with common air pollutants and other
35	environmental factors including UV and absolute humidity.
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37	A number of earlier studies on the environmental drivers of influenza transmission
38	used absolute counts of influenza cases as the dependent variable in statistical models.
39	However, the number of incident influenza cases is not an ideal representation of
40	influenza transmission intensity [3]. We estimated the daily effective reproduction
41	number (R_t), a real-time measure of transmissibility, for each influenza type/subtype
42	using data from the subtropical city of Hong Kong which has excellent influenza
43	surveillance data, near year-round circulation of influenza, and considerable variations
44	in environmental factors and pollutant levels. We combined information on influenza-
45	like illnesses in the community and laboratory surveillance data to estimate weekly
46	incidence rates of influenza virus infections in the community, referred to as ILI+ rates
47	[4] In theory this time series should be a linear correlate of the incidence rate of
19	infactions in the community [4] and it was proviously shown that there was a very close
40	hiere the community [4], and it was previously shown that there was a very close
49	correlation between this measure and laboratory confirmed H1N1pdm09
50	hospitalizations in Hong Kong in 2009-10 [5]. Finally, we multiplied the weekly ILI+
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51	rates by a large constant, representing the inverse of the coverage of the sentinel sites in
52	Hong Kong, and rounded to the nearest integer to obtain a time series of weekly ILI+
53	counts (Figure 1a-d). This was then interpolated to daily ILI+ counts using splines.
54	During the study period of January 1998 through December 2013, we identified 44
55	distinct influenza epidemics, including 16 epidemics of seasonal influenza A(H3N2), 10
56	of A(H1N1), 4 of A(H1N1pdm09), and 14 of influenza B (Figure 1a-d). Daily
57	concentrations of major air pollutants in 10 local monitoring stations were used to
58	calculate the territory-wide daily average concentrations for Hong Kong.
59	Meteorological data were obtained from the Hong Kong Observatory.
60	
61	Transmissibility can be measured by the effective (or instantaneous) reproduction
62	number (R_t) as an unit-free index of outbreak intensity, defined as the average number
63	of secondary infections caused by a typical single infectious person at time t , in the
64	population. We estimated R_t from daily ILI+ counts for each influenza type/sub-type.
65	We adopted a simple branching process model [6] to estimate daily R_t values. We
66	assumed a Gamma distribution for the serial interval with mean values of 3.08
67	(SD=1.39) for influenza A(H1N1pdm09), 3.26 (SD=1.93) for A(H1N1), 3.48 (SD=1.88)
68	for A(H3N2) and 3.72 (SD=1.95) for influenza B [7].
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70	We used regression models to explore the association between influenza
71	transmissibility, measured by the daily estimated effective reproductive numbers (R_t)
72	for up to 8 weeks either side of each epidemic peak, and various pollutant factors with
73	0-7 days lag values. In non-linear univariate regression analysis, we found that R_t had
74	statistically significant negative association with ambient O_3 across all the types/sub-

types; NO and CO had a weak positive association with influenza transmissibility, while 75

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70	other pollutants had no consistent patterns and the estimated effects were generally not
7	7 statistically significant. The estimated non-linear effect of ozone on influenza
78	3 transmissibility is shown in Figure 1f. The multivariable regression (DLM, dlmn package
79	in R) model that included depletion of susceptibles, inter-epidemic factors, absolute
8	humidity and ambient ozone could explain 40% of the observed variation in R_t for
8	1 seasonal influenza A(H3N2), 35% for seasonal influenza A(H1N1), 60% for
82	A(H1N1)pdm09 and 21% for influenza B. With a large proportion of the variance
83	8 explained by the intrinsic factors and absolute humidity in the basic model for influenza
84	transmissibility, the ambient ozone contributed only marginally, explaining a further
8	5 4% of the total variance in influenza transmissibility for H3N2 and up to 1% for the
8	other three influenza types/subtypes. A permutation analysis indicated that the
82	association was not likely to be due to chance (data not shown). While the proportion of
88	3 variance in influenza transmissibility explained by ozone is modest, this could still
89	9 correspond to a substantial effect on incidence in a single epidemic which includes
9(many transmission events [8]. In Hong Kong, seasonal influenza often exhibits twice-
9	annual peaks in periods from July to August (summer) and from January to March (late-
92	2 winter/early-spring) which generally coincide with two troughs of ozone concentration
93	3 seasonality (Figure 1a-e).

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95 The association of ambient ozone with reduced influenza transmissibility may be 96 related to ozone's virucidal activity and the effect of ozone on the host defense. Ozone 97 inactivation of influenza virus within a few hours has been reported in studies *in vitro* 98 [9]. However, a more plausible mechanism underlying the association of ozone with a 99 reduction in influenza transmissibility is ozone-primed immunity against influenza 100 virus infection. Inhalation of ambient ozone can enhance pulmonary innate immunity

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101	that promote allergic responses in healthy human subjects and susceptible populations
102	[10]. It is not likely for ozone as an oxidant gas to be directly recognized by a discrete
103	receptor; ozone-induced inflammation is probably mediated by a secondary messenger.
104	One such candidate is IL-33. Induced by ozone exposure, IL-33 further activates type 2
105	cytokines in the lung. IL-33 appears to be the common denominator for the list of
106	as thma triggers including allergy, viral infection, and O_3 [11]. As a multifaceted cytokine,
107	however, IL-33 plays not just a pathogenic role in Th-2 mediated diseases but also
108	drives TH 1 and CD8 T cell responses that induce protective immunity against viral
109	infections [12]. In the case of influenza, IL-33 promotes lung tissue homeostasis during
110	viral infection [13]. Used as an adjuvant in influenza vaccines, IL-33 increases the Ag-
111	specific CD4 and CD8 T cell responses in preclinical settings [14].
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113	One limitation of the present study was the interpolation of daily ILI+ counts from the
114	weekly data. The day-to-day variation in transmissibility might have been reduced
115	because of this interpolation, leading to underestimated effects of the drivers for
116	influenza. If available, using ILI+ data at a daily scale would improve the estimates.
117	Another limitation is that the territory-wide daily average calculation might introduce
118	measurement errors for certain pollutants such as NO_2 and CO which have a relatively
119	large spatial variability. However, if the spatial variability did not change systematically
120	with time, the aggregated exposure measurement should not bias the study findings
121	based on territory-wide time-series data of both influenza and environmental drivers.
122	As a highly reactive oxidant air pollutant, O_3 may decrease host defenses against
123	bacterial and fungal infections in the airways and aggravate pre-existing diseases such
124	as asthma. In the case of influenza, however, ambient O_3 had not been consistently
125	associated with hospital admissions or emergency department visits for influenza virus

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2 3 4	126	infections according to the review by the United States Environmental Protection
5	127	Agency in 2013 [10]. Our current findings of reduced influenza transmissibility
7 8	128	associated with ambient ozone in Hong Kong warrants further study.
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 - 184 BJC received research funding from Sanofi Pasteur for a study of influenza vaccine
 - 185 effectiveness.

- 187 Author contributions
- 188 BJC and LT designed the study. STA, PW, VJF and LT collected data. STA analysed data.
- 189 STA wrote the first draft, and all authors contributed to review and revision and have
- 190 seen and approved the final version.

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195	FIGURE LEGEND
196	Figure 1: (a-e) weekly activity of influenza (ILI+ proxy) by virus type/subtype (black
197	lines) along with the 44 predefined epidemics (gray bars), and the weekly smoothed
198	average of ozone concentrations in Hong Kong from 1998 through 2013. (f) Estimated
199	nonlinear relationship between the effective reproduction number R_t and ambient daily
200	ozone concentrations in the regression analysis (based on the selected best-fitting lag of
201	5 days for A(H3N2), 6 days for A(H1N1), 7 days for A(H1N1)pdm09 and 4 days for
202	influenza B) for influenza A(H3N2), A(H1N1) prior to 2009, A(H1N1)pdm09 from 2009
203	onwards, and influenza B. The violin plot shown in the lower panel indicates the
204	distribution of daily ambient ozone concentrations; the median is indicated by the white
205	circle, the interquartile range is indicated by the black rectangle, and the blue area
206	displays a kernel density estimate of the distribution of values (i.e. a smoothed
207	histogram).
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