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# **A Prospective Study of Ventilation Rates and Illness Absence in California Office Buildings**

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**Environmental Energy Technologies Division** 

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# <span id="page-5-0"></span>**Abstract**

*Background* – This study investigated the associations of ventilation rates (VRs), estimated from indoor  $CO<sub>2</sub>$  concentrations, in offices with the amount of respiratory infections, illness absences, and building-related health symptoms in occupants.

*Methods* – Office buildings were recruited from three California climate zones. In one or more study spaces within each building, real-time logging sensors measured carbon dioxide, temperature, and relative humidity for one year. Ventilation rates were estimated using daily peak  $CO<sub>2</sub>$  levels, and also using an alternative metric. Data on occupants and health outcomes were collected through web-based surveys every three months. Multivariate models were used to assess relationships between metrics of ventilation rate or

 $CO<sub>2</sub>$  and occupant outcomes. For all outcomes, negative associations were hypothesized with VR metrics, and positive associations with  $CO<sub>2</sub>$  metrics.

*Results –* Difficulty recruiting buildings and low survey response limited sample size and study power. In 16 studied spaces within 9 office buildings, VRs were uniformly high over the year, from twice to over nine times the California office VR standard (7 L/s or 15 cfm per person). VR and  $CO<sub>2</sub>$  metrics had no statistically significant relationships with occupant outcomes, except for a small significantly positive association of the alternative VR metric with respiratory illness-related absence, contrary to hypotheses.

*Conclusions–* The very high time-averaged VRs in the California office buildings studied presumably resulted from "economizer cycles" bringing in large volumes of outdoor air; however, in almost all buildings even the estimated minimum VRs supplied (without the economizer) substantially exceeded the minimum required VR. These high VRs may explain the absence of hypothesized relationships with occupant outcomes. Among uniformly high VRs, little variation in contaminant concentration and occupant effects would be expected. These findings may provide initial evidence for an upper bound of the range of VRs within which increased VRs provide benefits in reducing illness absence.

**Keywords** (3-10)

Ventilation rate, IAQ, SBS, illness absence

# <span id="page-6-0"></span>**Background**

The primary goal of this study was to provide quantitative estimates of the association between ventilation rates (VRs) in offices and adverse effects on building occupants – primarily respiratory illnesses and illness-related absences from work, but also acute health symptoms at work and dissatisfaction with the air quality at work. The results are intended to help support evidence-based, energy efficient, but health-protective ventilation standards for commercial buildings.

Standards for minimum VRs in commercial buildings historically have been based on laboratory studies of acceptability of air quality that considered occupants to be the only pollutant sources. More recently, standards have considered, to a limited extent, research on how VRs affect prevalence of "sick building syndrome" (SBS) symptoms. Research now suggests that current ventilation standards provide neither adequate protection from SBS symptoms nor satisfactory perceived air quality in offices [\[1-5\]](#page-29-1).

Pollutants in office buildings, which may cause adverse effects in occupants, can be emitted by the buildings and everything within, including furniture, equipment, and occupants themselves. Outdoor air brought into offices by mechanical ventilation systems is the primary means by which levels of indoor-generated pollutants are reduced. Heating or cooling this outdoor air to comfortable indoor levels requires increased energy as VRs increase. In setting energy-conscious VR standards, adverse effects on occupants from inadequate ventilation can be considered as costs to be weighed against the benefits of reduced energy use and energy costs. The human outcomes of potential concern in setting commercial VR standards (although not all are considered currently) include building-related symptoms, infectious respiratory disease, asthma exacerbations, illness-related work absence, reduced work performance, and poor perceived air quality.

Building occupants can emit infectious respiratory agents that cause illness in other occupants [\[6\]](#page-29-2). The hypothesis underlying this study is that sufficiently lower VRs in office buildings, indicated by higher measured carbon dioxide  $(CO<sub>2</sub>)$  concentrations, would lead to greater indoor air concentrations of agents causing infectious respiratory disease, which would lead to higher rates of illness absence in the occupants. This hypothesis is supported by a finding in a prior study [\[7\]](#page-29-3) of a 35% reduction in short-term illness absence among office workers with VRs of 24 L/s per person compared to those with 12 L/s per person, based on VRs estimated from  $CO<sub>2</sub>$ data. Milton et al. [\[7\]](#page-29-3) presumed that short-term illness absence was primarily from respiratory infections. Other prior findings [\[6\]](#page-29-2) also provide support for the hypothesis. There are currently few published data in the archival literature documenting associations between ventilation rate and illness absence in office environments.

Chemical and non-infectious biological pollutants indoors may also cause irritation, allergies, or dissatisfaction with indoor air quality. SBS symptoms have been used extensively as a measure of health-related outcomes in offices. Lower VRs have been associated with elevated prevalence and intensity of SBS symptoms [\[1,](#page-29-1) [8\]](#page-29-4). It is not known if these symptoms are severe enough to contribute to illness-related absence. SBS symptoms also could be considered as costs, to be weighed against energy benefits of lower VRs.

This study estimates VRs from measured indoor  $CO<sub>2</sub>$  concentrations, collected and logged in real time continuously for one year, minus estimated outdoor concentration. Also, since  $CO_2$  is a product of human respiration, the indoor  $CO<sub>2</sub>$  concentration itself can be considered as a proxy, in evaluating effectiveness of ventilation for controlling airborne concentrations of humanproduced infectious respiratory agents that contribute to absence. The study analyses estimate the associations of both VRs and  $CO<sub>2</sub>$  concentrations with occupant outcomes, including respiratory infections, illness absence, symptom severity, and perceived air quality. This information on human health effects will provide input into decisions about costs and benefits of decreasing or increasing minimum VR standards.

Many buildings have economizer control systems that increase VRs above a set minimum value, by increasing outdoor air flow rates during times of cool to moderate outdoor air temperatures, and reducing building energy consumption for air-conditioning. In general, minimum VRs are provided when the outdoor temperature is either above the desired indoor temperature or below approximately  $10^{\circ}$ C; however, control strategies vary somewhat from building to building. In much of California, economizers increase VRs above the minimum VR most of the time. Dutton and Fisk [\[9\]](#page-29-5) estimated that overall, for California offices with economizers, VRs will exceed the set minimum VR approximately 80% of the time.

# <span id="page-7-0"></span>**Methods**

#### *Building recruitment*

Buildings in California were solicited for participation by emails, flyers, and phone calls to the employers. Eligible office buildings were from the public or private sector in three distinct climatic regions of California – Bay Area, Central Valley, and South Coast. In each selected building, a study space with at least 30 occupants was selected, either a subset of the building and its workers or the full building, within which relatively uniform VRs were anticipated (e.g., contiguous space, space with air recirculation by air handling systems). A single building could contain multiple separate study spaces. Buildings or study spaces containing unusual contaminant sources were excluded. The target size of the study was a total of 30-40 study spaces, including 50 or more workers in each.

Given the high expected refusal rate during building recruitment (based on our prior experience), the sample was not intended to be representative of California commercial buildings, but a sample of convenience using available opportunities. Recruitment, enrollment, and data collection were conducted in a rolling manner, so that data collection began in the earliest recruited buildings while other buildings were still being recruited. Data were collected for at least a full year within each building, but study periods were not simultaneous across all study buildings.

#### *Environmental Data*

Several types of environmental data were collected: measurements of indoor  $CO<sub>2</sub>$ , temperature (T), and relative humidity (RH), along with selected characteristics of the buildings and ventilation systems. Other indoor air pollutants were not measured.  $CO<sub>2</sub>$  was monitored by Vaisala CARBOCAP<sup>TM</sup> #GMW110 sensors (Vaisala Inc., Boulder CO). HOBO T & RH loggers (Onset Computer Corporation, Cape Code, MA) were used to measure T and RH and to log the  $CO<sub>2</sub>$  data.

CO2, T, and RH were measured at continuous 10-minute intervals at 2-3 indoor locations per study area. In an initial visit at each building, the sensor packages  $(CO<sub>2</sub>$  sensor plus T and RH data loggers) were installed at suitable locations away from likely direct occupant exhalation, e.g. attached to the top of space partitions and in common areas such as hallways. A contact at each building was queried about which 2-hour period in the morning in each study space was most likely to have a stable number of occupants. Each quarter-year, in-place sensor packages were replaced with sensor packages containing newly-calibrated  $CO<sub>2</sub>$  sensors. Sequential waves of replacement were used within the study buildings. Seventy sensor packages were used in the study, 66 more than once.

Data from the two to three sensors within each study space were first averaged at each time point to provide overall real-time estimates for the study space. As the primary VR metric for analysis (VR Method 1), real-time spatially averaged  $CO_2$  data from 8 a.m. to 5:30 p.m. on workdays were used to estimate daily workday VRs (as outdoor airflow rates in L/s per person) using the equilibrium  $CO_2$  method; i.e., from observed peak moving 60-minute-averaged  $CO_2$ concentrations, per ASTM D6245-12 [\[10\]](#page-29-6). These calculations assumed that daily VRs in each office were stable,  $CO_2$  reached equilibrium daily in each. The outdoor air flow Q (m<sup>3</sup>/h) was estimated from the maximum indoor  $CO<sub>2</sub>$  concentration as follows:

$$
\frac{Q}{N} = \frac{S}{\left(C_{\text{max}} - C_o\right)} \left(\frac{h}{3600 \text{ s}}\right) \tag{1}
$$

where Q/N (L/s-person) is the per-person outdoor airflow rate,  $C_o$  (g/m<sup>3</sup>) is the outdoor  $CO_2$ concentration (set to 380 ppm here),  $C_{max}(g/m^3)$  is the maximum hourly averaged  $CO_2$ concentration measured indoors between 8:00 a.m. and 5:30 p.m., and S is the  $CO_2$  generation rate, set at 18.6 L/h-person (Mudarri, 1997) for sedentary persons with an activity level of 1.2 met units.

The underlying assumptions of equation one were often violated, leading to inaccurate VRs. An alternative method (VR Method 2) was also used for estimating VRs, based on the build-up of indoor  $CO<sub>2</sub>$ , required no assumption about equilibrium but assumed stable ventilation rates and occupancy during selected periods. This method considers build-up of  $CO<sub>2</sub>$  during selected 2hour periods in each study space with relatively stable occupancy numbers, and also in the afternoon after workers returned from lunch. Additional information about the  $CO<sub>2</sub>$ measurements and about both VR estimation methods is available in Additional file 1.

These two estimators of daily VR were intended for use in two ways: to calculate prior 3-month averages of daily VRs, for analyses with the occupant outcomes involving occupant recall over the prior 3 months (respiratory infections and illness absences); and as daily values, for analyses with the occupant outcomes linked to the day of the occupant survey (symptoms and perceived air quality). Metrics of both daily time-averaged mean  $CO<sub>2</sub>$  and daily peak 60-minute-averaged

 $CO<sub>2</sub>$  concentrations were also used in analyses without conversion to VR estimates, as both 3month-average and daily values.

The particle filtration in the heating, ventilating, and air-conditioning (HVAC) systems in the study buildings were characterized via interviews with building managers and data obtained from filter manufacturers. Use of economizer cycles in the HVAC was determined by interviews with building managers.

#### *Human outcomes data*

Initial development of tools and procedures for data collection from occupants included a human subjects consent form, a web-based survey tool developed for administration via the Internet (based on revision of a previously used survey), and data handling protocols to ensure the confidentiality of personal information. Before collecting human subjects data, a human subjects protocol was approved by the Lawrence Berkeley National Laboratory Human Subjects Committee.

Data on occupants and their outcomes were obtained from occupant surveys every three months during the study, starting three months after initial sensor installation in the building, using the web-based survey tool. See Table 1 for a schedule of sensor installation and survey administration. In the initial survey for each participant only, data were obtained on personal/demographic variables that can influence risk of respiratory illness (age, gender, smoking status, asthma status), home variables (young children at home), and work factors (job type, office space sharing, hours worked in building). See Additional file 2 for questions in the initial and recurring surveys.

In the initial and in each recurring survey, data were obtained on the number of episodes of infectious respiratory illnesses and the number of days of absence caused by respiratory illnesses, in the prior three months. These surveys also included questions on perceived air quality and on severity of four symptoms on the day of the survey. Respondents were asked to rate the indoor air quality in two questions, on acceptability of air quality and acceptability of odor, with the response scale for each ranging from 1 to 7 ("clearly acceptable" to "clearly unacceptable"). The

**Table 1**. Schedule for sensor installation and survey administration in each building during the year of study



\* in period 1, initial installation of calibrated sensors; in later periods, replacement by newly calibrated sensors

symptom questions included: dry, itching, or irritated eyes; headaches; unusual tiredness or fatigue; and congested nose, asking respondent to rate the symptom at that time on a severity scale from 0-10 (none to very severe) and also, if they reported that symptom at a level of 1 or higher, whether they had the symptom before arriving at work that day.

To improve survey response rate, a small financial incentive was provided automatically when each survey was completed: upon submission of each survey, a \$4 gift certificate was provided in an email as a numeric code for online redemption, except that for the fourth survey among those also completing the prior three surveys, an \$8 incentive was provided. In several buildings, financial incentives were not allowed. In one of these buildings, however, the facility manager instituted a competition between the two study spaces there on their survey response rate throughout the study. The building managers received no information on responses of individuals.

#### *Analysis*

Collected survey data were omitted from occupants who reported working less than 20 hours per week in their building and from those who failed to complete an initial questionnaire with background information.

Environmental data, collected in real time during the study period, were excluded from analyses outside the weekday hours of 8:00 a.m. – 5:30 p.m., on U.S. federal holidays, and during periods of local shutdown at university buildings. Also, any day in a study space with no apparent elevation of indoor  $CO_2$  above approximately 400 ppm was excluded as a non-work day in that space.

Data collected were analyzed to assess relationships between estimated ventilation rates or  $CO<sub>2</sub>$ concentrations, either daily or averaged over the prior 3-month periods, and occupant outcomes assessed in the survey at the end of each quarter (Table 1). Data analyses were performed using Stata v. 11 (StataCorp LP, College Station, TX, USA; www.stata.com). Analyses of the longitudinal data were at the individual (subject) level, and included unadjusted and adjusted models accounting for repeated measurements on individuals.

Analyses provided point estimates and confidence intervals for the estimated relationships between variables. Appropriate statistical models were selected for analysis of each type of human outcome, all using "bootstrap" procedures for variance of estimates to account for clustering on individuals and study spaces. For respiratory illness episodes and illness absence days, zero inflated negative binomial, zero inflated Poisson, negative binomial, or Poisson models were used, which produce point estimates of incident rate ratios (IRRs). (Details of how these models were chosen are provided in Additional file 3.) For symptom outcomes, which have highly skewed distributions with many zero values, zero-inflated negative binomial models were used. For perceived air quality and odors, generalized estimating equation (GEE) linear and logistic regression models were used, which produce point estimates of regression coefficients and of odds ratios (ORs), respectively.

All adjusted models included covariates for potential confounding as appropriate. For repeated measures analyses within individual subjects and study spaces, adjustment for unchanging personal variables as potential confounding was not necessary. For analyses of respiratory illness episodes and related absences, a covariate was included in models for a "respiratory illness season." (Plots of prior respiratory illness by month showed higher numbers reported on surveys in the months of January through April for illness in the prior three months, corresponding to a season of increased respiratory illness spanning October-April; this was used to define the respiratory illness season.) Models for each symptom severity outcome on the day of the survey included a covariate indicating whether the respondent already had that specific symptom when arriving at work. The models used, along with the specific types of exposure variables (e.g., estimated ventilation rate or  $CO<sub>2</sub>$  concentrations) and the covariates included in models, are described in Additional file 3.

### <span id="page-11-0"></span>**Results**

Building recruitment was challenging: only a small proportion of contacted buildings agreed to participate. A total of 17 separate study areas within 10 office buildings were successfully recruited for participation. Due to loss of environmental data, 16 study spaces in nine buildings were included in analyses (Table 2). Two included spaces contained fewer than 50 office workers. All the included buildings but two (engineering firms) were in the public sector (state or municipal government, higher education, research). All study spaces had air-conditioning and were reported to have economizers. Data collection from sensors in the first participating building began in May 2012, and the first occupant survey was conducted in that building 3 months later, in August 2012. (Data collection was continuous except in study spaces 2a and 2b, where a major furniture move after the third period required a 3-month suspension of the study before proceeding with the final 3 month period.) Completed data collection from sensors and surveys was concluded in all study spaces by October, 2013, except in space 9, which was enrolled so late that data from the fourth survey was not available in time for analysis deadlines .



Table 2. Buildings participating in the HZEB Office Building Ventilation Rate Study



 $*$  end date of  $3<sup>rd</sup>$  survey,  $4<sup>th</sup>$  survey not included in this space

Abbreviations: MERV*,* Minimum Efficiency Reporting Value; N, number; occs, occupants.

#### *Occupant data*

Response rates to the occupant survey were lower than expected, despite use of financial incentives (Additional file 4).: the 1,297 valid surveys received represented an overall 27% response on the four surveys, varying from 16 to 41% across study spaces. Response rates for individual surveys in each study space ranged from 8 to 54%. However, the incentives, of about \$4-\$8 for each 5-minute survey, did increase response over the non-incentive study spaces by 50% (from 18% to 27%). The competition set up between two nonincentive study spaces within one building (1a and 1b), with no prize other than pride in winning, produced a response rate 78% higher than other regular non-incentive spaces, and even 18% higher than the spaces with financial incentives.

Table 3 provides information on the study respondents. No data were available to allow comparison of survey participants to nonparticipants. Respondents included slightly more males (53%), included a broad range of ages from under 30 (17%) to over 50 (29%), and were highly educated (98% with at least a college degree, 45% with a graduate degree). Most (81%) had no children up to age 3 years at home. Most (78%) reported never smoking. Half (50%) reported some history of allergy or asthma, including 25% for hay fever and 16% for asthma, and 11% reported current asthma. Most participants (75%) worked in open office spaces; 70% shared their workspace with at least 7 others, with only 18% in private offices. Over half (54%) reported working over 40 hours per week. Most (68%) reported high levels of job stress, but only 26% reported high levels of job dissatisfaction.



# Table 3. Characteristics of survey respondents\*

\* after exclusion of surveys from workers who worked <21 hours/week in the building and from those not completing an initial survey with background data \*\* from initial survey response to a question repeated on each survey \*\*\* proportions are calculated using total of non-missing answers

Table 4 shows the distributions, in each study space and overall, for the number of respiratory infection episodes reported in the prior three months and for respiratory illnessrelated absences in the prior three months. For the number of respiratory infection episodes in the prior 3 months, the overall mean was 0.92, with a range across study spaces from 0.67 to 1.32. The 95<sup>th</sup> percentile value overall was 3, ranging in specific study spaces from 2 to 4. For the number of respiratory illness-related work absences in the prior 3 months, the overall mean was 0.78, with a range across study spaces from 0.10 to 1.38. The 95<sup>th</sup> percentile value overall was 4, ranging in specific study spaces from 1 to 6.



Table 4. Respiratory illness outcomes among respondents\*

\* after all exclusions (see Table 3 footnote)

Table 5 describes symptoms reported on the days of the surveys, in each space and overall, including all eligible surveys. The overall proportions of respondents reporting any of eye, headache, fatigue, or nose symptoms at work (considering all eligible surveys from all study spaces together) were 65%, 35%, 61%, and 51% respectively, with respective mean severity scores among those reporting any of each symptom of 4.4, 3.8, 4.4, and 4.0 out of 10. For the eye, headache, fatigue, and nose symptoms, the minimum proportions reported in any study space were 40%, 15%, 44%, and 30%, and the maximum proportions reported were 82%, 57%, 74%, and 68%, respectively (excluding study space 9, which had few total responses). Among those reporting any symptom responses for eye, headache, fatigue, and nose symptoms, the ranges across study spaces of symptom severity scores were 3.0-5.4, 2.3-5.1, 3.6- 6.2, and 3.1-4.8. Additional file 5 shows the number and proportion of occupants in each building with prior experience of each type of symptom on the day of the survey. Prior symptoms (before work on the survey day) were common: overall proportions of surveys reporting prior eye, headache, fatigues, and nose symptoms were 38%, 31%, 56%, and 65%, respectively







\* symptoms are: dry, itching, or irritated eyes; headaches; unusual tiredness or fatigue; and congested nose; response scale ranges from 0 (none) to 10 (very severe) \*\* after all exclusions (see Table 3 footnote); includes data from respondents eligible for survey and who also answered both parts of the symptom question

Table 6 describes the reported acceptability of indoor air quality and odors, by space and overall, including all eligible surveys. The proportion of surveys rating the indoor air quality as unacceptable (on a dichotomous scale) was 10.2% overall, ranging across study spaces from 2.5 to 30%. Averaged over the 4 surveys in each space, only two of 16 spaces (8c and 9) failed to provide acceptable air quality for at least the minimum 80% proportion of occupants assumed in the American Society of Heating, Refrigerating, and Airconditioning Engineers (ASHRAE) 62.1 ventilation standard. The overall mean IAQ acceptability score for all surveys was 4.6 (on a continuous scale, with 1=barely acceptable and 10=completely acceptable), ranging across study spaces from 3.0 to 6.1. The proportion of surveys rating the odors as unacceptable was 3.7% overall, ranging across study spaces from 0 to 11%. The overall mean odor acceptability score for all surveys was 5.5, ranging across study spaces from 4.1 to 6.3.

#### *Environmental data –*

Based on recalibration of  $CO<sub>2</sub>$  sensors after deployment in the field for 3-month periods, sensor drift was small, approximately  $\pm 5\%$  (see Additional file 1). Temperature sensors were determined to have read fairly consistently 1 °C high, due to an internal heat source in monitoring modules and temperature values used in modeling were adjusted accordingly.

Table 7 summarizes, by study space and specific survey periods for each, median values for the prior three months of three VR-related variables: daily VRs (Method 1), daily mean  $CO<sub>2</sub>$ , and daily maximum  $CO<sub>2</sub>$ . Quarterly (per three-month) median VRs in study spaces ranged from 6.9 to 65.8 L/s per person, medians of daily mean  $CO_2$  from 425-957 ppm, and medians of daily maximum  $CO_2$  from 494-1230 ppm. VRs were uniformly high relative to the current minimum VR standards for office space: 8.5 L/s (17 cfm) per person from ASHRAE 62.1, at the default density of occupancy, and 7 L/s (15 cfm) per person from California Title 24. Other than one median quarterly VR of 6.9 L/s-person in space 4, all other quarterly medians exceeded 13 L/s per person, or almost double the California



Table 6. Environmental acceptability outcomes

standard. Figure 1 shows distributions of daily maximum  $CO<sub>2</sub>$  measurements, over the entire study, by study space. Figure 1 shows that the study spaces had generally low maximum  $CO<sub>2</sub>$ concentrations and thus high VRs, except space 4, which had a slightly higher  $CO<sub>2</sub>$  distribution. Relative  $CO_2$  levels across buildings were similar for maximum and mean  $CO_2$ . Distributions of daily mean  $CO_2$  values are provided in Additional file 6. Figure 2 shows daily maximum  $CO_2$ values over time in each study space. Most study spaces had relatively uniform maximum  $CO<sub>2</sub>$ throughout the study, with the exception of space 7 and, to a lesser extent, 8b. Patterns for maximum and mean daily  $CO<sub>2</sub>$  over time were similar. Space 4, Survey 3, had the most low VRs and high  $CO<sub>2</sub>$  levels; otherwise the ranges across buildings and surveys were narrower. Distributions of daily mean values over time are provided in Additional file 7.

Additional file 8 shows the distributions of VRs in each study space estimated using VR Method 1, i.e., calculated with equation 1, plus the  $5<sup>th</sup>$  percentile values as indicators of minimum VRs. Minimum VRs based on the  $2.5<sup>th</sup>$  percentile were similar. The estimated minimum VRs

<b>Study Space</b>	<b>Three-Month Median of</b> <b>Daily VRs:</b> $(L/s$ -person)				Three-Month Median of Daily Mean CO2: (ppm)				Three-Month Median of Daily Maximum CO2: (ppm)			
	S <sub>1</sub>	S <sub>2</sub>	S <sub>3</sub>	<b>S4</b>	S <sub>1</sub>	S <sub>2</sub>	S <sub>3</sub>	<b>S4</b>	S <sub>1</sub>	S <sub>2</sub>	S <sub>3</sub>	<b>S4</b>
Bay Area												
1a	21.2	19.2	19.7	20.9	571	605	607	58	669	703	702	679
1 <sub>b</sub>	18.8	17.1	17.1	17.6	$\sqrt{573}$	$\overline{603}$	$\sqrt{603}$	$\overline{579}$	$\overline{723}$	731	$\overline{749}$	$\overline{729}$
2a	21.3	17.9	13.9	19.6	569	615	660	587	661	713	803	678
2 <sub>b</sub>	21.5	18.4	13.8	22.0	570	602	657	566	660	699	803	660
3a	17.8	25.2	22.1	19.8	550	534	513	533	713	628	664	672
$\overline{3b}$	38.9	36.4	31.3	39.4	470	482	481	463	538	547	578	541
3c	13.7	24.4	19.6	15.4	587	532	526	534	873	652	704	843
3d	29.8	34.9	$\overline{3}1.9$	28.6	480	496	475	483	588	571	566	602
6	65.8	41.9	32.6	38.2	432	471	492	477	500	547	580	567
Central Valley												
$\overline{4}$	18.8	13.1	6.9	14.0	602	663	957	656	720	825	1230	822
9	20.7	25.6	20.3	<b>NA</b>	563	529	577	<b>NA</b>	733	699	703	<b>NA</b>
South Coast												
5a	15.4	17.5	16.0	15.8	646	605	647	653	759	732	777	764
$\overline{5b}$	17.6	20.1	17.2	20.9	555	572	580	569	711	682	752	673
7	23.1	28.4	27.0	27.1	574	530	538	541	659	594	618	629
8 <sub>b</sub>	27.9	23.4	58.1	52.6	512	507	446	441	726	835	568	584
8c	30.8	41.6	58.8	64.9	525	471	448	425	581	528	501	494

Table 7. Prior three-month\* median of daily ventilation rates (VRs)\*\*, daily mean CO<sub>2</sub>, and daily maximum CO<sub>2</sub>, by study space

\* prior three-month period ending on the first day of each survey period in each space

\*\* VR Method 1



Figure 1. Distributions of daily maximum indoor CO<sub>2</sub> measurements, by study space grouped by climate zone (boxes show median,  $25<sup>th</sup>$  and  $75<sup>th</sup>$  percentiles; whiskers,  $75<sup>th</sup>$  percentile plus 1.5 times the interquartile distance, and  $25<sup>th</sup>$  percentile minus 1.5 times the interquartile distance)



Figure 2. Daily CO<sub>2</sub> maximum indoor values over time (y-axis, in ppm, per study space grouped by climate zone

substantially exceeded the 7 L s<sup>-1</sup> per person requirement in most study spaces, with an average estimated minimum of 15 L/s per person. In 13 of 16 study spaces, the estimated minimum VR exceeded 10 L/s per person.

Because the estimated daily VR metric was extremely variable, analyses used 3-month VR averages. Measured  $CO<sub>2</sub>$  metrics but not estimated daily VRs were used in analyses of same-day symptoms or acceptability of air. Because economizer systems were present in all study spaces it was not possible to include economizer presence in models. The efficiency of particle filters, expressed as a Minimum Efficiency Reporting Value (MERV) rating ranging from 1-16, was clustered at two values, 8 and 14 MERV, with the higher efficiency present in the study spaces in only two buildings, so it was also not possible to include this in models.

#### *Environment and outcome results*

Additional file 9 provides summaries of occupant outcomes by categories of various demographic and personal variables of occupants. Workers in private offices had the lowest proportions of respiratory infections and days of respiratory illness-related work absence, and the highest scores for acceptability of indoor air quality (IAQ) and odors, but these outcomes did not worsen consistently as the number of others sharing the workspace increased. Sharing workspace with fewer others also did not show consistent associations with fewer symptoms. Very low job stress and low job dissatisfaction were associated with unusually low levels of respiratory illness-related absence, relatively high acceptability of IAQ and odors, and somewhat lower proportions reporting most symptoms. Smokers reported relatively low levels of respiratory illness episodes and related work absence, relatively low levels of most symptoms, and higher levels of environmental acceptability. Females reported many more respiratory illness-related work absences, more of most symptoms, and somewhat lower environmental acceptability.

Table 8 summarizes the associations, unadjusted and adjusted, between  $CO<sub>2</sub>$  and VRs in the prior three months and the two respiratory illness outcomes, estimated from zero-inflated negative binomial models. Covariates and their categories used in these adjusted models are described in Table 3. None of the unadjusted or adjusted estimates were significantly associated with  $CO<sub>2</sub>$  or VR metrics. All estimates not equal to 1.0 were in directions opposite those hypothesized (below rather than above 1.0 for the  $CO<sub>2</sub>$  metrics, and above rather than below 1.0 for VR). For analyses using the alternative VR metric based on curve-fitting for  $CO<sub>2</sub>$  increases, results were similar, except that for respiratory illness-related absences, with increased VR there was a small statistically significant increase in estimated illness absence– OR (95% CI) 1.015 (1.0005-1.03),  $p=0.043$  – but not in respiratory illness episodes – 1.001 (0.90-1.008),  $p=0.78$ .

Additional file 10 summarizes associations between the same variables but estimated from logistic regression models. All the unadjusted estimates were in the hypothesized directions, and all the adjusted estimates in directions opposite those hypothesized, although none were statistically significant. The directions of all adjusted estimates, and the magnitudes for the respiratory illness episodes, were similar to (or showed smaller effects) those from the zeroinflated negative binomial models.

Of the four other covariates in models for the two respiratory illness-related outcomes, only respiratory illness season had strong and consistent associations, with highly significant IRRs of about 1.5 and, from logistic regression models, ORs of 2.0 for both illness episodes and days of illness absence. Shared workspace had highly significant positive associations only in the logistic models, with ORs of 1.6 for illness episodes and 2.0 for days of illness absence. Young children at home had significantly elevated associations only in logistic models for illness episodes, with ORs of 1.7. Ever smoking had no consistent associations with either outcome.

Table 8. Unadjusted and adjusted associations, as incident rate ratios (IRRs) and 95% CIs,\* between  $CO<sub>2</sub>$  and ventilation rates in the prior three months, and respiratory illness outcomes, estimated from zero inflated negative binomial (or, as noted, negative binomial or zero-inflated Poisson) models



\* The IRR is interpreted as the multiplicative change in estimated rate of outcomes for each increase of 100 ppm  $CO<sub>2</sub>$  or 1 L/s per person of VR. Estimates for VR models were hypothesized to be in the opposite direction as  $CO<sub>2</sub>$  models.

^ Negative binomial model

# Zero-inflated Poisson model

## Models adjusted for: smoking, young children in home, people sharing workspace, respiratory illness season (illness reporting period in October—April); see Table 5.

Table 9 summarizes the associations between same-day  $CO<sub>2</sub>$  metrics and symptoms (as continuous outcomes), estimated from zero-inflated negative binomial models. None of the adjusted estimates were statistically significant, and most were small (changes of 1% or less) and in the direction opposite that hypothesized. Table 10 summarizes the associations between the same-day  $CO<sub>2</sub>$  metrics and acceptability of indoor air and odor (as dichotomous outcomes), estimated from logistic regression models. (The ORs in this table are for the IAQ or odor being



Table 9. Unadjusted and adjusted<sup>#</sup> associations between same day  $CO_2$  and symptoms (as continuous outcomes), estimated as incidence rate ratios (IRRs) from zero inflated negative binomial models

100 ppm) # Models adjusted for job dissatisfaction, age, gender, education, smoking, shared workspace, any allergic history, and current asthma (see Table 5 for covariates) and also prior symptom before work and mean indoor temperature on day of survey.

Table 10. Unadjusted and adjusted<sup>#</sup> associations, estimated as odds ratios (ORs), between sameday  $CO<sub>2</sub>$  and acceptability of indoor air (as dichotomous outcomes), estimated from logistic regression models



# Models adjusted for job dissatisfaction, age, gender, smoking, and shared workspace (see Table 5 for covariates) and also mean indoor temperature on day of survey.

judged acceptable vs. unacceptable.) None of the adjusted estimates were statistically significant, and directions were mixed.

Table 11 summarizes the associations between the same-day  $CO<sub>2</sub>$  metrics and acceptability of indoor air and odor (as continuous outcomes), estimated from linear regression models. (The coefficients in this table are for the additive change in acceptability score for IAQ or for odor, on a scale of 1-7, with positive coefficients indicating improved acceptability.) None of the adjusted estimates were significant, and magnitudes were mixed. Female gender and greater job dissatisfaction were associated with less acceptability of IAQ.

Table 11. Unadjusted and adjusted<sup>#</sup> associations between same-day  $CO<sub>2</sub>$  and acceptability of indoor air (as continuous outcomes), estimated from linear regression models



# Models adjusted for job dissatisfaction, age, gender, smoking, and shared workspace (see Table 5 for covariates) and also mean indoor temperature on day of survey.

### <span id="page-24-0"></span>**Discussion**

The objective of this study was to quantify the relationships of VRs in California office buildings with occupant outcomes that were hypothesized, based on prior research, to be increased by lower VRs: respiratory illnesses and respiratory illness-related absences, building-related symptoms, and dissatisfaction with indoor air quality and odors. No statistically significant relationships were found, except for a small significantly positive association of the alternative VR metric and respiratory illness-related absence, contrary to hypotheses. Given that over 35 associations were estimated, one or two statistically significant association would have been expected simply by chance, without true underlying relationships. Some nonsignificant tendencies, such as for the  $CO<sub>2</sub>$  metrics and acceptability of odors measured on a continuous scale, were in the direction hypothesized from prior knowledge. In contrast, however, some nonsignificant tendencies, such as for the  $CO<sub>2</sub>$  metrics and the illness absence-related outcomes, were in directions opposite those hypothesized from limited prior findings.

The overall weakness of these signals suggest that actual relationships, within the range of VRs included in this study, were either absent or so weak that greater statistical power would be necessary to detect them. VRs were uniformly high over time in almost all study spaces. For the three-month median VRs in each study space, used in illness absence analyses, only one (6.9 L/s per person) was below 13; others ranged from 13.1-65.8 L/s per person. Most VR data in illness absence analyses (between the tenth and ninetieth percentiles) were between 16 and 42 L/s per person, which is over twice to over nine times the California minimum VR standard. For the daily VR values corresponding to the  $CO<sub>2</sub>$  values used in analyses of symptoms and perceived air quality, the tenth-to-ninetieth percentile range was 13 to 45 L/s per person. Thus this study was unable to assess relationships with VRs considered substandard, and could only compare high with very high VRs, a range in which indoor contaminant levels are highly diluted and little variation in contaminant concentration would result. The findings might in fact be interpreted as preliminary evidence for an upper bound of the range of VRs within which increased VRs may substantially reduce illness absence (about 16 L/s per person), or improve symptoms or perceived air quality (about 13 L/s per person). Establishing such bounds more firmly would require larger studies.

Many reviews have concluded that lower building VRs are associated with adverse human outcomes [\[1-3,](#page-29-1) [8,](#page-29-4) [11-14\]](#page-29-7). The limited available data suggest increases in absence rates and respiratory illnesses as ventilation rates are reduced [\[6,](#page-29-2) [7\]](#page-29-3). Milton et al. reported lower rates of illness absence in offices with 24 vs. 12 L/s per person, a contrast within the VR range of the current study. In studies of classrooms, Mendell et al. [\[15\]](#page-30-0) found that lower VRs in the range of approximately 2-20 L/s per person were associated with significantly increased illness absence in primary school students. A large number of office studies generally shows worsening of SBS symptoms and perceived air quality at lower VRs below about 20 or  $25 \text{ Ls}^{-1}$  per person [\[1-3\]](#page-29-1). This study did not find such relationships, but daily VRs in this study were mostly above  $13 \text{ L s}^{-1}$ per person.

#### *Strengths and limitations*

Strengths of this study include the prospective design, following office workers over four seasons during a full year, which allowed within-person analyses and reduced statistical confounding by personal factors; use of daily VRs or  $CO<sub>2</sub>$  in each study space based on real-time CO<sup>2</sup> measurements every day over a year instead of the usual short-term measurements over one or several days; and use of frequently recalibrated  $CO<sub>2</sub>$  sensors to estimate VRs, successfully keeping sensor drift within 5%.

The study had multiple limitations. A primary limitation was the insufficient statistical power, with a study size too small to detect the small differences in exposure and effects expected within the observed range of VR. The sample size was smaller than planned, a combined result of the inability to recruit the desired number of buildings (due to the unwillingness to participate of management in most buildings contacted), and the very low survey participation rates of occupants in participating buildings, despite financial incentives. Findings may apply only to public-sector buildings, as most contacted in the private sector declined to participate, and to the minority of occupants who participated in each study space. Respiratory illness episodes and respiratory illness-related work absence were assessed only by questionnaire, and retrospectively for the prior three month period. Prospective gathering of this data from occupants would have been more accurate, but also more onerous and susceptible to nonresponse.

Estimation of VRs involved many potential sources of error, as assumptions underlying use of equation 1 are often invalid: peak  $CO<sub>2</sub>$  levels in each space not reaching true equilibrium during many work days (resulting in overestimation of true VRs as well as random error); potential errors in measuring and estimating indoor  $CO<sub>2</sub>$  levels in each study space, such as from poor air mixing or nearby occupant exhalation (resulting in underestimation of VRs); the use of a fixed rather than measured outdoor  $CO<sub>2</sub>$  levels, which vary by location and time of day, in calculating VRs (resulting in random VR errors); possibly inaccurate assumptions about  $CO<sub>2</sub>$  generation rate; and the assumption of unchanging VR per person during each day in each space, despite occupancy changes, and part-day use of economizers. VRs calculated from the alternative metric rely on fewer assumptions, but only represent VRs during the selected short time periods.

#### *Implications*

Ventilation rate standards are still largely based on decades-old studies of the amount of ventilation needed to satisfy 80% of visitors to a space with the occupants as the dominant pollutant source [\[16,](#page-30-1) [17\]](#page-30-2). There is no explicit analysis underlying the current standards that considers health risks from exposure to indoor air pollutants, potential impacts to workers' performance, or energy and other associated cost considerations. Further research is still necessary to provide scientific support for health-protective building VR standards.

The uniformly high VRs in the California office buildings in this study are presumably due largely to the combination of generally moderate climates and use of economizer systems that bring in large volumes of outdoor air (for "free cooling") during periods of moderate outdoor temperatures. However, even the estimated minimum VRs observed in these buildings (when presumably operating without economizers) also generally exceeded the minimum requirement in the California Title 24 standard, suggesting poor control of minimum VRs. (Some proportion of these high VR estimates was, as discussed previously, likely due to overestimation based on daily peak  $CO<sub>2</sub>$  levels.) Future assessments of the relationships studied in this project thus need to include geographic areas with more severe climates and thus lower VRs. Future studies should also take into account the possibility that in some locations very high VRs may introduce substantial amounts of outdoor air pollutants into the indoor environments, with adverse respiratory effects.

The conduct and findings of the present study also provide other lessons for future studies on this topic. Substantial time and effort will be necessary to successfully recruit a sufficiently large sample of buildings. Within study buildings, either substantially increased financial incentives or other novel approaches will be necessary to achieve desired response rates, especially in a prospective study with repeated surveys. Use of prospectively collected diary data from occupants on respiratory infections, or even more objectively, employer-provided illness absence data, would improve the quality of these data. Improved methods to estimate VRs need to be developed and validated. (Although tracer gas methods can measure VRs more accurately, they are not practical for year-long studies in multiple buildings.) Increased introduction of outdoor air pollutants into buildings by higher VRs should be considered in data collection and analyses, especially with higher VRs. To the extent that specific indoor office pollutants that might vary with VR or even without VR could influence respiratory illness, measurement of these over time would reduce statistical noise and allow greater power in smaller studies; however, such measurements could be quite costly.

The prospective observational design used in the present study, with some of the improvements mentioned above, would be suitable for the questions of interest here, as it is relatively economical and it allows greater generalizability of findings than controlled chamber studies, which also could not study respiratory infections in office populations over extended periods. However, field intervention studies comparing existing low and experimentally raised VRs, if done for extended periods in consideration of seasonal illness patterns, and in large populations to achieve sufficient power, could provide additional useful information on VR and respiratory infections.

In addition, developing increased knowledge about physical mechanisms of indoor transmission of respiratory infections would help focus future field studies on determinants of transmission. It is still uncertain how much of this disease transmission occurs by each of four possible modes: a direct contact mode (person-to-person contact); an indirect contact mode (from physical contact with surfaces contacted by those infected); a droplet mode (the impact of large droplets from coughing and sneezing on others quite nearby); or by long-range airborne transmission (through very small droplet aerosols produced by drying of larger aerosols expelled in coughing or sneezing). The VR may influence transmission of infectious respiratory disease by changing indoor concentrations of the very small aerosols associated with long-range airborne transmission or possibly by changing humidity or concentrations of air pollutants that might affect either the period of viability of infectious particles or people's susceptibility to infection. However, all of the disease transmission mechanisms are coupled, e.g., increased long-range airborne transmission would result in more sick occupants who could transmit infections via other transmission mechanisms. A number of infectious agents are involved in the mix of

infectious diseases in an office population, and each may be primarily transmitted through different mechanisms. Improved understanding of these processes will help inform field research in buildings.

# <span id="page-27-0"></span>**Conclusions**

This study found (with one exception) no statistically significant relationships between VRs in these California commercial buildings and occupant outcomes hypothesized to be increased by lower VRs: respiratory illnesses and respiratory illness-related absences, building-related symptoms, and dissatisfaction with indoor air quality and odors. (The exception was one small, significantly positive association of the alternative VR metric with respiratory illness-related absence, contrary to hypotheses, but a possible chance finding due to the large number of analyses.) The overall lack of relationships was apparently due to the almost uniformly very high VRs in the studied spaces over the year of the study. The three-month median VRs in the study spaces, with one exception, ranged from 13.1 to 65.8 L/s per person, which is from almost twice to over nine times the California minimum VR standard of 7 L/s (15 cfm) per person. Thus this study had limited contrast in exposures, and could compare only high with very high VRs, a range in which little variation in contaminant concentration and occupant effects would result.

This study provided some limited data on actual minimum VRs in California office buildings (during non-use of economizers), suggesting that, to the extent the studied buildings are representative, these VRs are usually substantially higher than required in the current applicable standard. This conclusion is limited by potential errors of overestimation for VRs in this study.

#### **List of abbreviations**



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# <span id="page-31-0"></span>**Additional Files**

#### **Additional file 1. Primary and alternate methods used for estimating daily ventilation rates**

#### **Introduction**

The Healthy Zero Energy Buildings (HZEB) study on ventilation rate (VR) and worker illness absence collected  $CO_2$  concentrations for one year in 16 office spaces in California. The 16 study spaces are located in nine office buildings. This document describes the procedures used to estimate VRs from the real-time indoor  $CO<sub>2</sub>$  data measured at 10-minute intervals.

Over the course of a workday,  $CO<sub>2</sub>$  concentrations typically increase as workers arrive at the office. There is usually a brief drop in concentrations during lunchtime. After that,  $CO<sub>2</sub>$ concentrations tend to increase again as workers return to their desks. At the end of the workday,  $CO<sub>2</sub>$  concentrations decrease again as workers leave the office. This varying occupancy throughout the workday is one of the challenges of estimating  $VR$  from  $CO<sub>2</sub>$  data. Two methods are used to estimate the ventilation rates from the  $CO<sub>2</sub>$  concentrations measured. Ventilation rates are estimated in units of air changes per hour, and also in terms of outdoor airflow rate per person (liters of air per second per person).

#### **Data Collection**

In each of the study spaces, concentrations of  $CO<sub>2</sub>$  were monitored using the Vaisala CARBOCAP® carbon dioxide transmitter GMW110 sensor, each was mounted on a wall or hung on a partition within the space. This device uses an infrared absorption sensor measuring at the CO<sup>2</sup> absorbance wavelength. It also uses a microchemical Fabry-Perot Interferometer (FPI) filter to make a reference measurement where no absorption occurs. This allows the device to compensate for potential light intensity variations, dirt accumulation, and other interferences in the optical path. The reported long-term stability of the GMW110 is  $+/-5\%$  of range (0 to 2000) ppm) over 5 years. The sensor also has negligible dependence on temperature (-0.35% of reading per  $^{\circ}$ C) and pressure (+0.15% of reading per hPa). The device has a response time of 1 minute.

In this study, data were logged every 10 minutes at two or three locations within each study space. After three months, each sensor was replaced with a newly calibrated sensor for the next three months. The GMW110 outputs a voltage between 0 and 20 mA, which is logged using HOBO voltage data loggers. A calibration curve was used to compute the  $CO<sub>2</sub>$  concentrations from the recorded voltage. This calibration curve was obtained once before a sensor was deployed, and again three months later after the sensor finished sampling. Calibration was performed by placing the Vaisala  $CO_2$  sensors in a small room, where  $CO_2$  was injected to raise the concentrations to between 1000 and 2000 ppm.  $CO<sub>2</sub>$  concentration then decayed gradually over approximately 12 to 24 hours to background levels. The decay in concentrations was measured by a PPSystems  $CO<sub>2</sub>$  gas analyzer EGM-4, which is a high precision instrument with reported accuracy of <1% over the concentration span between 0 and 2000 ppm. The calibration curve was obtained by a best-fitted regression line that describes the relationship between the voltage output by the Vaisala and the  $CO<sub>2</sub>$  concentrations measured by the PPS ystems gas analyzer.

Each Vaisala  $CO<sub>2</sub>$  sensor was used up to five times in this study in different study spaces. In between uses, the sensors were calibrated as described above. Seventy sensors were used in this study, of which 66 were used more than once. [Figure 1](#page-32-0) compares the differences in  $CO<sub>2</sub>$ concentrations between calibrations from the 66 sensors. Changes between calibrations were evaluated by comparing the  $CO<sub>2</sub>$  concentrations that a sensor reports now relative to the values that it would report using the prior calibration curve. [Figure 1](#page-32-0) shows that the Vaisala  $CO<sub>2</sub>$ sensors tend to report higher values after use than when the prior calibration curve was used. This is mainly caused by a tendency towards positive offsets, as shown in boxplot (a). On the other hand, the slope centers on unity, meaning that on average, sensitivity of the sensors to  $CO<sub>2</sub>$ concentrations remained unchanged. Dirt accumulation absorbing some of the infrared signal may be one of the causes of positive offset. The resulting change is about 10 ppm (median value) if  $CO<sub>2</sub>$  concentrations were at 600 ppm, as shown in [Figure 1\(](#page-32-0)c). But performance was worse in some of the sensors, where a gain of 40 ppm was observed in 10% of the comparisons.



<span id="page-32-0"></span>**Figure 1 Change in CO<sup>2</sup> readings measured by the Vaisala sensors between calibrations. This analysis included a total of 177 comparisons from 66 sensors that were used more than once in this study.** 

#### **Methods**

 $CO<sub>2</sub>$  concentrations measured by multiple sensors were averaged to better represent the indoor concentrations of the study space. Two methods are considered in this analysis to estimate VR from this spatially averaged  $CO<sub>2</sub>$  data.

**Method 1** assumes that indoor  $CO_2$  concentrations measured each day reached steady state such that the outdoor air flow Q ( $m<sup>3</sup>/h$ ) is determined from the maximum  $CO<sub>2</sub>$  concentration measured indoors as follows:

$$
\frac{G}{Q} = (C_{\text{max}} - C_o)Y \quad \text{where} \quad Y = \frac{0.183 \frac{g}{m^3}}{100 \text{ ppm}} \tag{1}
$$

where G (g/h) is the total CO<sub>2</sub> generation rate from all occupants, Q (m<sup>3</sup>/h) is the outdoor air flow, and  $C_0$  (ppm) is the outdoor  $CO_2$  concentration.  $C_{\text{max}}$  (ppm) is the maximum hourly averaged  $CO<sub>2</sub>$  concentration (that is, the maximum value of a moving 60-minute average) measured indoors between 8 am and 5:30 pm.  $C_0$  is set to equal 380 ppm in all the analyses presented here. A  $CO_2$  generation rate,  $S = 18.6$  L/h-person (Mudarri, 1997) is used to estimate the VR per person. This  $CO<sub>2</sub>$  generation rate corresponds to sedentary persons with an activity level of 1.2 met units (see ASHRAE Standards 62.2-2012 Appendix C). Substituting  $G = N \times S$ into equation (1), where N is the number of occupants, gives:

$$
\frac{Q}{N} = \frac{S}{\left(C_{\text{max}} - C_o\right)} \left(\frac{h}{3600 \text{ s}}\right) \tag{2}
$$

where Q/N (L/s-person) is the per-person outdoor airflow rate.

**Method 2** makes use of  $CO_2$  data from a period when the number of office workers is roughly stable. This occurs typically in the morning, when most workers have already arrived at work and before lunchtime. During this stable period, the rate of increase in indoor  $CO_2$  is reflective of the ventilation rate per occupants, and also the air change per hour  $Q/V$  (h<sup>-1</sup>).

If the indoor air is assumed to be well mixed, then the governing equation for indoor  $CO<sub>2</sub>$ concentration, C, is as follows:

$$
V\frac{dC}{dt} = Q(C_o - C) + \frac{G}{Y}
$$
\n(3)

$$
C = C_o + \frac{G}{QY} + \frac{\mathfrak{E}}{\mathfrak{E}} C(t_o) - \frac{G}{QY} - C_o \frac{\mathfrak{E}}{\mathfrak{E}} e^{-\frac{Q}{V}t}
$$
(4)

where  $C(t_0)$  is the initial indoor  $CO_2$  concentration, V is the building volume (m<sup>3</sup>), and  $C_0$ , Q, and G are as defined in equation (1).

Equation (4) is solved using a nonlinear least squares fitting function in R statistical software. The function uses the Gauss-Newton algorithm, which is an iterative procedure, to solve for the values of G/Q and Q/V. A solution is accepted if it met the tolerance level for convergence  $(10^{-5})$ within the maximum number of iterations allowed (50). The starting estimate for G/Q is computed from equation (1), and  $Q/V = 0.5$  h<sup>-1</sup> is used in all cases as the initial guess.

The per-person VR and occupant density per floor area can be roughly estimated from the fitted values of G/Q and Q/V, as follows:

$$
\frac{Q}{N} = \frac{SY}{G'Q} \tag{5}
$$

$$
\frac{N}{FA} = \frac{N}{V/H} = \frac{\left(\frac{Q}{V}\right)H}{\frac{Q}{N}} = \frac{\left(\frac{G}{Q}\right)\left(\frac{Q}{V}\right)H}{S} \cdot \frac{100 \text{ m}^2}{Y}
$$
(6)

In equation (6), the factor of 100 is for computing the occupant density, N/FA, in units of number of people per 100  $m^2$  of floor area. A typical office ceiling height of 2.75 m (9 ft) is assumed in equation (6).

There are many cases where Method 2 did not result in reasonable estimates of G/Q and Q/V, even though the convergence was achieved. Acceptability criteria is defined as p-value of the fitted Q/V parameter being less than 0.05. For this analysis, this effectively rejected unrealistic estimates of G/Q and Q/V that would result from poor fitting of the data. Each study space provided a two-hour period where occupancy was reported to be stable, typically between 9 and 11 am. In two of the sixteen study spaces (B3S2 and B3S4), it was necessary to shift the stable period from the reported times of between 9:30 and 11:30 to half an hour earlier, such that Method 2 would give reasonable estimates of ventilation rates for at least some of the days.

Equation (4) describes a steady increase in indoor  $CO<sub>2</sub>$  concentrations as a function of time during a stable occupancy period. It also assumes that the outdoor air flow is constant. To screen out days where the measured  $CO<sub>2</sub>$  did not fit these descriptions, the nonlinear parameter fitting was only performed if the  $CO<sub>2</sub>$  concentration measured towards the end of the two-hour stable period was at least 20 ppm higher than the beginning.

In some of the study spaces, there were substantial differences in the morning and afternoon  $CO<sub>2</sub>$ peak concentrations. Such morning-versus-afternoon differences are often more pronounced in certain seasons, suggesting that this likely resulted from the economizer bringing in more outdoor air for cooling when outdoor conditions were favorable. When a rise in  $CO<sub>2</sub>$ concentrations was observed in the afternoon, equation (4) was used to obtain another set of G/Q and Q/V parameters for that period using the same procedure as described above for the morning period. It is assumed that most people returned from lunch by 1:30 pm, and most people remained in the office until 4:30 pm. Based on this assumption, the afternoon stable period is set to start when the  $CO<sub>2</sub>$  concentration was the lowest between 1:30 and 2 pm, and end when the  $CO<sub>2</sub>$  concentration was the highest between 4 and 4:30 pm.

When estimates of G/Q and Q/V were successfully obtained for both morning and afternoon, estimates from the two periods were averaged to give a daily estimate. If Method 2 only gave acceptable estimates for either the morning or afternoon, G/Q of the remaining period was estimated using Method 1. For example, if Method 2 successfully estimated the values of G/Q and Q/V for the morning but not the afternoon, the afternoon G/Q was estimated using equation (1), and the per-person outdoor airflow rate, Q/N, was estimated by equation (5). The air changes per hour can also be approximated by assuming that occupant density was about the same before and after lunch. In this example, Q/V for the afternoon period can be estimated using equation (6) and assuming that the value of N/FA that was determined from the morning period also applied for the afternoon. This procedure allows an estimate of Q/N and Q/V for the afternoon period even when Method 2 failed to converge, so that a daily estimate can be computed by taking the average of the two periods.

#### **Additional file 2. Occupant survey questionnaires**

 Note: All questions in the below initial survey were repeated in each recurring survey, except questions 8.1 through 8.5, which were asked only in the initial survey.

#### **Important – Please complete this survey at** *your usual workstation at work***, and answer the questions about** *the indoor environment there***. It will probably take you less than 5 minutes. We would like you to answer all the questions. However, you can skip a question if you do not want to answer it. To skip a question, check "***no answer***" and go to the next question.**







 **For the following symptom,** please choose the appropriate response: **Mark the circle that represents how severe this symptom is for you at the CURRENT TIME.** 3(a) **How severe is this symptom for you now: headache? Very Severe None 1 2 3 4 5 6 7 8 9 10**  *no answer*

 [*This question appears only for those who answered 1-10 for part a*] 3(b) **Did you have this symptom before you arrived at work today?**

<br> $123$ Please choose ONE response:  $\qquad$   $\qquad$ 

- No
	-
- *no answer*



 151 **For the following symptom,** please choose the appropriate response:<br>152 **Mark the circle that represents how severe this symptom is for you** Mark the circle that represents how severe this symptom is for you *at the CURRENT TIME*. 154<br>155

 5(a) **How severe is this symptom for you now: congested nose?** 





 **6.3. In the last 3 months, on how many days were you absent from work (for a whole day)** *because of these respiratory illnesses***? Please report as well as you can remember.**











#### 321 **Additional file 3. Description of analysis models, with exposure variables and covariates** 321<br>322





323  $*$  maximum sliding 15-minute average  $CO<sub>2</sub>$  over the workday hours of 830 a.m.-530 p.m.

324 \*\* includes only symptoms beginning at work

325 \*\*\* if illness reporting period (3-month period prior to survey) within October—April

 $326$  \*\*\*\*in NB model component; in ZI model component included only  $CO_2$ , season, number people in work area, and hours worked per week people in work area, and hours worked per week

 $^{7+++}_{7++++}$  Primary models estimate VR from peak daily CO<sub>2;</sub> secondary models estimate VR from curve-fitting algorithm; both described in Additional file 1. curve-fitting algorithm; both described in Additional file 1.

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# 333 **Additional file 4.** Occupant responses





<sup>335</sup> \* response proportion for each survey calculated as proportion of original total of occupants

336 occupants<br>337 \*\* after ex

<sup>337</sup> \*\* after exclusion of surveys from occupants working in the building only 20 or fewer<br><sup>338</sup> hours/week, and not completing the initial survey with background information

hours/week, and not completing the initial survey with background information

339



#### 341 **Additional file 5**. Prior symptom\* outcomes among respondents on day of survey\*\*

342

<sup>343</sup> \* symptoms are: dry, itching, or irritated eyes; headaches; unusual tiredness or fatigue; and congested nose; response scale ranges from 0 (none) to 10 (very severe)

344 congested nose; response scale ranges from 0 (none) to 10 (very severe)<br>345 \*\* after all exclusions (see Additional File 4 footnote); includes data from

\*\* after all exclusions (see Additional File 4 footnote); includes data from respondents eligible

346 for survey and who also answered both parts of the symptom question

347

348

350<br>351

**Additional file 6.** Distributions of daily mean indoor CO2 measurements, by study space grouped by climate zone

- grouped by climate zone
- 



 



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- 
- 





**Additional file 8. Distributions of ventilation rates calculated via method 1, and estimated minimum ventilation rates from 5th** 364 365 **percentile values.**



# **Additional file 9**. Outcome summaries by categories of respondent characteristics



372<br>373 373 Note: All numbers exclude those not working >20 hours/week

374 \* after exclusion of those with a prior symptom

- <sup>375</sup> \*\* For these variables, questions were asked only on the first survey, and responses were retained for analyses of later surveys <br><sup>376</sup> + with or without cubicles or partitions
- $376$  + with or without cubicles or partitions<br> $377$  + pollen allergy
- ++ pollen allergy

Additional file 10. Adjusted and unadjusted associations between CO<sub>2</sub> and ventilation rates in the prior three months and respiratory illness outcomes, estimated from logistic regression models)



# Models adjusted for: smoking, young children in home, people sharing workspace, respiratory illness season (illness reporting period in October—April); see Table 5.