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2020

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Citation Details

Gall, E. T., Mishra, A. K., Li, J., Schiavon, S., & Laguerre, A. (2020). Impact of Cognitive Tasks on CO2 and Isoprene Emissions from Humans. Environmental Science & Technology.

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1 Impact of cognitive tasks on $CO₂$ and isoprene ² emissions from humans

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11 **1. Abstract**

- 12 The human body emits a wide range of chemicals, including $CO₂$ and isoprene. To examine
- 13 the impact of cognitive tasks on human emission rates of $CO₂$ and isoprene, we conducted an
- 14 across subjects, counterbalanced study in a controlled chamber involving 16 adults. The
- 15 chamber replicated an office environment. In groups of four, participants engaged in 30
- 16 minutes each of cognitive tasks (stressed activity) and watching nature documentaries (relaxed
- 17 activity). Measured biomarkers indicated higher stress levels were achieved during the
- 18 stressed activity. Per-person CO₂ emission rates were greater for stressed than relaxed activity
- 19 $(30.3 \pm 2.1 \text{ vs. } 27.0 \pm 1.7 \text{ g/h/p}, p = 0.0044 \text{, mean } \pm \text{ standard deviation}).$ Isoprene emission
- 20 rates were also elevated under stressed vs. relaxed activity ($154 \pm 25 \,\mu g/h/p$ vs. 116 ± 20 21 μ g/h/p, $p = 0.041$). Chamber temperature was held constant at 26.2 ± 0.49 °C; incidental
- 22 variation in temperature did not explain variance in emission rates. Isoprene emission rates
- 23 increased linearly with salivary-alpha amylase levels ($r^2 = 0.6$, $p = 0.02$). These results imply

24 the possibility of considering cognitive tasks when determining building ventilation rates.

25 They also present the possibility of monitoring indicators of cognitive tasks of occupants

- 26 through measurement of air quality.
- 27 **Keywords:** human emissions; bioeffluents; stress biomarkers; CO2; isoprene

28 **2. Introduction**

- 29 Human chemical emissions of gaseous or particle-phase compounds associated with
- 30 human metabolism include carbon dioxide,¹ volatile organic compounds (VOCs),² and
- $bioaerosols^{3,4} These compounds, emitted via human breath and skin, impact indoor air$
- 32 chemistry and contribute to degraded indoor air quality.^{5,6} Human chemical emissions, often
- 33 indicated by proxy measurement of carbon dioxide $(CO₂)$, are also a driver of the need to
- 34 ventilate buildings, with prominent building ventilation standards like ASHRAE 62.1 and EN

16798 based on removing odorous human chemical emissions from the indoor environment.⁷ 35 Exposure to human chemical emissions may also impact human cognition.^{8,9} Studies have 37 implicated exposure to pure elevated CO_2 as impairing human cognition,^{10,11} though there also 38 exist studies that show no effect of CO_2 itself on cognition.^{12,13} However, it is consistently 39 shown that reduced outdoor air ventilation, leading to higher indoor concentration of human 40 chemical emissions, is responsible for observed decrements in cognition or environment 41 perception. $14-16$

42 A variety of factors beyond ventilation rate and occupant density impact the level of 43 human chemical emissions in a given indoor space. There are methods for predicting the $CO₂$ 44 generation rate per-occupant, $17,18$ and they consider body surface area, body composition, 45 metabolic rate (related to type of activity), air temperature, and a respiratory quotient, which is 46 Iargely a function of diet.¹⁹ There are many studies in the literature quantifying type and quantity of emissions of endogenous and exogenous volatile organic compounds, $20-23$ often in 48 pursuit of understanding the human "volatilome" for diagnosis of disease.²⁴ Enabled by 49 analytical methods that can measure VOCs with high time resolution, studies have estimated 50 average, per-person emission rates of carbon dioxide and VOCs in a variety of field settings 51 including classrooms,^{2,25,26} a museum,²⁷ and a movie theater²⁸; these studies have substantially 52 expanded our knowledge of the magnitudes and type of human chemical emissions. However, 53 comparatively fewer studies investigate factors in healthy humans that may influence human 54 chemical emissions, including traits like age, sex, or smoking status, $29-31$ and their 55 biochemical mechanisms.³² Recent studies, conducted in controlled chambers, measured 56 human emissions under varying "human factors" including type of clothing worn and age, as 57 well as environmental factors, including temperature. $33,34$

 The impact of psychological factors on human chemical emissions is not well-studied 59 or understood. In a 1975 field study of human bioeffluent emissions, Wang et al.³⁵ measured volatile emissions from students in a University classroom used for lectures and examinations. They found twelve organic compounds were elevated during lecture periods, and estimate per- person production rates. In that study, the researchers identified and separated examination periods as a condition where occupants experienced increased stress relative to lecture. They 64 reported a 43% increase in $CO₂$ emission rate when comparing examination to lecture conditions; the authors suggest that this was due to elevated metabolism. More recently, 66 Williams et al.³⁶, investigated the role of audiovisual stimuli to influence human chemical emissions using a high time-resolution proton transfer reaction - time of flight - mass spectrometer installed in the heating, ventilation and air-conditioning system of a movie theater. By annotating films with scene content labels, the researchers ascertained whether a scene label could predict variation of the measured human chemical emissions. Carbon dioxide and isoprene exhibited high correlations with scene labels, indicating emissions of these compounds may be associated with biochemical pathways induced by emotional responses to the movie scene. Isoprene emissions and variability have also been studied as a 74 potential marker for age-appropriateness of film content,³⁷ and may serve as an indicator for several metabolic processes in the human body, including lipid metabolism and physiological 76 state.

 Better understanding of factors influencing human chemical emission, including stimuli such as cognitive tasks and audiovisual experience, will advance our understanding of metabolic and biochemical pathways that may be initiated as a result of cognitive processes. Such knowledge could be applied to improve ventilation strategies in buildings by, for

 example, tailoring ventilation strategies to the psychological experience and/or commitments of the occupants. Conversely, monitoring levels of human chemical emissions in buildings could be used to gauge occupant cognitive engagement. To date, studies advancing our knowledge of factors influencing human chemical emissions have generally occurred in uncontrolled field environments (movie theaters, classrooms, museums). We use a state-of- the-art controlled test chamber to engage sixteen subjects in an across subjects, counterbalanced study, to explore how cognitive tasks may influence human chemical emission rates of isoprene and carbon dioxide. To our knowledge, this is the first controlled study of the effect of cognitive tasks on human chemical emission rates.

3. Methods and materials

3.1 Test chamber and subjects

 We conducted studies in a room-scale test chamber to measure the influence of 93 cognitive tasks on $CO₂$ and isoprene emitted by humans. The test chamber is located within the Singapore-Berkeley Building Efficiency for Sustainability in the Tropics facility (Figure S1, Supporting Information). The test bed consists of four climatic chambers, each 5.6×4.3×2.7 m. Our tests were conducted in one of the four chambers mocked up as a typical office environment with desks and computer workstations. Four participants and one researcher were present during test periods, shown in Figure 1 (a) and (b). An experimental timeline, post-occupancy by test subjects, is shown in Figure 1 (c). The test chamber was served by a single-pass, mixing ventilation forced air heating, ventilation, and air-conditioning (HVAC) system that includes supply air cleaning, with MERV8 and pleated activated carbon (International Filtration, Carbon Pleated Interfold). Supply air is drawn into the chamber HVAC system from the surrounding indoor space, which is a laboratory environment. A ceiling fan with transparent blades (diameter = 1.47 m, Artemis Minka-Aire F803L-TL model,

105 Minka Group, USA) was kept operational during the study sessions to improve mixing of the 106 chamber air with low distractions. Experimental and outdoor conditions during each test are 107 shown in Table S1 and Table S2 of the Supporting Information.

108
109

Figure 1. Overview of experimental set-up (a) and (b) and time-line of experiments post-110 occupancy (c). $SF = \text{subjective feedback}$, $ETC = \text{eye tracker calibration}$, $SF-C = \text{subjective}$ 111 feedback & capnometer, $TCS =$ thermal comfort monitoring station, $CO_2 =$ Wall mounted 112 CO_2 sensor, NM = noise meter, Cap. = capnometer.

- 113
- 114 *3.2 Measurement of participant stress levels*

 Our test protocol recruited participants to engage in an experiment where human chemical emissions were measured while engaged in two activities: a "relaxed" activity where the participants viewed a 30-minute nature documentary (two clips of nature documentaries, 15 min each, the first a "tour" of redwood forests and the second a hummingbird documentary

 narrated by Sir David Attenborough) and a "stressed" activity where the participants engaged in a 30-min series of cognitive tasks. We selected five cognitive tasks covering the executive 121 functions: working memory, inhibition, and cognitive flexibility/task switching.³⁹ The tasks 122 were administered through the PEBL platform.⁴⁰ Prior studies suggest that audiovisual stimuli 123 can alter stress state, observable by biomarkers like cortisol and salivary alpha amylase, $41,42$ 124 with viewing of stressful or soothing videos of \sim 30 min in length.^{42,43} We include further details of cognitive tests used in the Supporting Information. In total, we tested sixteen participants in groups of four, with the order of exposure to the relaxed and stressed activities alternating over the four days of testing to achieve a counterbalanced design. Participant characteristics are shown in Table S3 of Supporting Information. The test protocol was approved by the University of California Berkeley Ethics Committee (Protocol #IRB2019-07-12403). Participants were briefed on the study in a one-hour familiarization session conducted for each participant prior to the experiments.

 We measured objective and subjective indicators of occupant stress and perceptions of the test-chamber's indoor environment. Participants arrived 15 minutes prior to the start of the session and were outfitted with a heart rate sensor (Polar H10, Polar, Finland). During the session, each participant sat at a workstation with a laptop. Each laptop had an eye tracker installed (Tobii Pro Nano, Tobii Pro, Sweden) to measure pupil dilation. Upon completion of the relaxed and stressed activities, we measured salivary alpha amylase (Cocoro Meter, Nipro 138 Corporation, Japan) and end tidal CO_2 (etCO₂) level (CO_2 monitor, OLG-3800, Nihon Kohden, Japan). We also collected subjective feedback concerning indoor environmental quality and mental effort; further details are provided in the Supporting Information. After the

session was complete, we measured participants' weight and body composition using a

bioelectrical impedance based monitor (RD-545IM, Tanita, Japan).

3.3 Human chemical emission measurement

 Human chemical emissions evaluated in this study include carbon dioxide and isoprene. Human chemical emissions were measured during a 30-min baseline period, occurring prior to the arrival of occupants, and during the two 30-min periods during occupancy, coinciding with relaxed or stressed activity (Figure 1 (c), baseline not shown). In 148 1-min intervals, measurements of $CO₂$ were made in the chamber supply air and in bulk air in the test chamber (GMW84, Vaisala, Finland, Range 0-2000 ppm, uncertainty: larger of 30 ppm and 3%). Time-integrated measurements of volatile organic compounds were made during baseline and occupied periods via sampling with sorbent tubes packed with 100 mg of 152 Tenax TA and 180 mg of Carbograph 1, based on a method developed previously,⁴⁴ modified for target compounds, sampling times, and sampling flowrate (~130 mL/min, total sample volume = 4 L). One portable sampling pump (PCXR4, SKC, USA) drew sample air through perfluoroalkoxy (PFA) tubing connected to the room supply air duct and another sampling 156 pump did the same from the bulk room air. Both sampling pumps had \sim 2 m of 0.635" OD tubing placed upstream of the sorbent cartridge. Bulk room air samples were made in duplicate using a flow splitter. Sorbent cartridges were analyzed with thermal desorption – gas chromatography – mass spectrometry (TD-GC-MS, TurboMatrix 650, PerkinElmer, USA connected to a gas chromatograph, Model 7890 A, Agilent Technologies, USA). We quantified VOCs by generating calibration curves from a calibration mixture that included isoprene and acetone. We do not report acetone levels or emission rates due to uncertainty in quantification, apparent in high variability across duplicate samples: the mean relative percent difference (RPD) across all duplicate isoprene concentrations measured was 10.9% while the

 RPD for acetone duplicates was 66.8%. Since acetone is emitted by humans with high source strength, we suspect the method is not suitable for robust acetone quantification at the levels present in the chamber. Further details of the sampling and analysis method are provided in the Supporting Information and in Figure S2. To return chamber bioeffluent levels to near- background levels, during the break between the two activities the room was flushed with a 170 fan (flowrate = $62 \text{ m}^3/\text{min}$, IBF300, Aman, India) providing ~10 turnovers of air during the 15-min break.

 Chamber temperature and relative humidity were measured continuously in 1-min intervals (ThermCondSys 5500, Sensor, Poland) and ozone levels were monitored (Model 205, 2BTech, USA) once per day, prior to the arrival of occupants, to ensure near-zero ozone levels; ozone was consistently <2 ppb, the stated uncertainty of the instrument. Light levels at each workstation ranged between 931 and 957 lux, which is within normal ranges of typical office work environments. The chamber windows were blacked out to ensure consistent lighting levels and avoid variations due to outside lights. We determined the air-exchange rate in the chamber from measurements of air flow rates in the supply duct as well as tracer decay tests conducted following the conclusion of the experiments for the day (excepting day 1) (see Table S1). The chamber is well-mixed, confirmed via measurements of temperature and 182 airflow gradients, as well as high r^2 (>0.99) when applying a well-mixed model to CO_2 tracer decay periods. The sorbent tubes and sampling pumps were located in a chamber adjacent to the experimental test chamber, and were not visible or audible to participants (instruments were in Chamber III, participants in Chamber IV, Figure S1 of Supporting Information).

3.4 Data analysis

3.4.1 Calculation of CO² emission rates

188 Time-series measurements of $CO₂$ were used to estimate emission rates of $CO₂$ for the 189 baseline and 30-minute periods during which participants were engaged in either relaxed or 190 stressed activity. We used a non-linear curve-fit applied to the analytical solution of a mass-191 balance equation that included air-exchange and $CO₂$ emissions from occupants, shown in 192 equation 1:

$$
C_{CO2,t} = \left(C_{CO2,0} + \frac{S_{CO2}}{\lambda V}\right)\left(1 - e^{-\lambda t}\right) + C_{CO2,t=0}e^{-\lambda t}
$$
\n(1)

193 where, $C_{CO2,t}$ is the concentration of CO_2 in the chamber at time *t* within the 30-min period of a 194 "relaxed" or "stressed" activity (g m⁻³), $C_{CO2,0}$ is the time-averaged concentration in supply air 195 over the 30-min period (g m⁻³), S_{CO2} is the chamber CO₂ source strength due to occupancy (g 196 h^{-1}), λ is the air-exchange rate due to ventilation of the chamber (h⁻¹), V is the volume of the 197 chamber (m³), *t* is time since the start of the 30-min test period (h), and $C_{CO2,t=0}$ is the CO₂ 198 level in the chamber at the start of the 30-min test period $(g m⁻³)$.

199 A best-fit value of S_{CO2} over the 30-min period was determined using the "fit" function 200 in MATLAB R2019a, with the fit type specified as eq. 1. The chamber was occupied by five 201 persons during all occupied test conditions; we report $CO₂$ source strengths on a per-person 202 basis by dividing the best-fit S_{CO2} from eq. 1 by five. Note that one of the occupants was 203 always the same researcher, present in the chamber as per IRB protocol requirement. He 204 maintained a similar diet, exercise level, and sleep schedule during the week of testing, and 205 endeavored to maintain similar during-test activity level across all tests.

206 *3.4.2 Calculation of isoprene emission rates*

207 Isoprene was measured over three 30-min periods: baseline (unoccupied), stressed and 208 relaxed activity (both occupied by five persons). Measured isoprene levels are time-integrated 209 over these 30-min periods. In the case of the baseline period, the chamber was previously

 unoccupied and operated at constant air-exchange for ~18 hours, and we assume the chamber has reached steady-state. For unoccupied conditions we calculate the isoprene emission rate according to equation 2:

 $S_{isoprene, baseline} = \lambda (C_{i, chamber} - C_{i,0})V$ (2) 213 where $S_{isoprene, baseline}$ is the source or sink strength of isoprene (μ g h⁻¹) under baseline 214 conditions, λ is the air-exchange rate due to ventilation of the chamber (h⁻¹), C_i is the 215 concentration of isoprene in chamber air (μ g m⁻³), $C_{i,o}$ is the concentration of isoprene in 216 supply air (μ g m⁻³) and *V* is the volume of the chamber (m³). Measurements of isoprene were corrected for the field blank, converted from mass to concentration using the sample volume for the specific sampling event. For periods of occupancy, we developed a method for estimating the isoprene source strength using a mass-balance model as in equation 1, written for isoprene. Because our measurement of isoprene was time-integrated, we integrated the mass-balance model over the 30-min period of relaxed or stressed condition, in 1-min time-step, for either relaxed or stressed activity period to determine the area under the concentration-time curve. Note that the relaxed or stressed condition began 10 min following occupancy (Figure 1c). This area calculation was established as the objective function, and the source strength of isoprene over the 30-min period was varied until the modeled area is equal to that of the concentration-time area known from the time-integrated measurement, as shown in equation 3:

^{40 min}
\n
$$
\sum_{t=10}^{40 min} C_{i, model, t} \times \Delta t_{model} = C_{i, measured} \times \Delta t_{measured}
$$
\n^{41 min}
\n⁴² where $C_{i, model, t}$ is the modeled chamber concentration of isoprene, calculated in 1-min
\n⁴³ intervals using a mass-balance as shown in equation S1 of the Supporting Information (µg m³), Δt_{model} is the time resolution of the discrete integration of the model concentration-time

231 curve (1 min), $C_{i,measured}$ is the time-integrated measurement of isoprene in the chamber,

232 taken as the average of duplicate measurements of isoprene and corrected for field blank (µg

233 m⁻³), and $\Delta t_{measured}$ is the time duration of the measured isoprene level in the chamber (30-min).

 A detailed example of the method employed for determining isoprene emission rates is provided in the supporting information in Table S4, with a graphical presentation Figure S3, present in the Supporting Information.

3.4.3 Statistical tests

 As shown in Table S1 of the Supporting Information, we conducted tests over four 240 days where CO₂ and isoprene source or sink strengths were measured for baseline, relaxed, and stressed activity, with order of relaxed and stressed activity alternating each day. To 242 determine if the test conditions resulted in statistically significant changes in $CO₂$ or isoprene source or sink strength in the chamber, we used analysis of variance (ANOVA) across three 244 groups (baseline, relaxed, stressed), similar to a method described previously.⁴⁵ Briefly, if the 245 three-group comparison resulted in α < 0.05, we then made direct comparisons across groups (i.e., baseline vs. relaxed, baseline vs. stressed, and relaxed vs. stressed) using Tukey's honest significant difference test (Tukey test). We included the baseline (unoccupied) conditions in ANOVA to evaluate whether source or sink strengths during occupied conditions were significantly different from unoccupied conditions. All statistical analyses were performed using MATLAB 2019a.

251 Measurements of biomarkers (HRV, SAA, pupil dilation, and end tidal $CO₂$) of stress were analyzed using either parametric (t-tests) or non-parametric tests (Wilcoxon rank sum tests) depending on if the data was normally distributed or not. Significance level was taken at

- 254 0.05 except for the HRV parameters, where, due to the large number of comparisons,
- Benjamini–Hochberg procedure (false discovery rate taken as 10%) was used. All
- comparisons were paired and two tailed.
- Data and analysis scripts/functions are publicly available and can be downloaded from
- the Dryad repository: https://doi.org/10.5061/dryad.gb5mkkwmk.
- **4. Results and Discussion**
- *4.1 Objective and subjective indicators of occupant stress*
- Objective measured stress biomarkers shown in Figure 2 indicate that the cognitive
- tests (stressed activity) induced a state of occupant stress compared to watching nature
- documentaries (relaxed activity). While most of the physiological parameters examined
- should increase with stress levels, the inverse is true for some parameters, specifically the
- Mean RR interval (the interval between heart beats) and fraction of power of heart beats in the
- lower frequency bands. To create a simpler visual presentation, we present these parameters
- multiplied by -1 for plots. This ensures that all the plotted parameters are low when stress is
- low and are high when stress is high.

 Salivary alpha amylase levels are elevated during stressed activity compared to relaxed 276 activity (*p*-value = 0.0014, $d = 0.75$, a medium to large size effect). Salivary alpha amylase levels are known to follow a diurnal course, and prior studies recommend controlling for time 278 of day;⁴⁶ our design had occupants entering the chamber at ~14:00 (Table S1) each day and the relaxed/stressed activities occurred over a period of ~120 min. Nevertheless, the study was counterbalanced, which will address the potential for confounding due to diurnal changes in salivary alpha amylase that may have occurred over this 120-min period, though we expect 282 this effect to be small based on the diurnal profiles presented by Nater et al.⁴⁶. We analyzed 283 51 parameters of heart rate variability; as noted from previous studies, not all parameters are expected to yield a significant difference when stress is imposed on people. Note that one participant's data was lost on day 3; subsequent analysis used data from fifteen participants. In

 Figure 2 are four metrics showing differences due to external stress, evident due to statistically significant differences from the relaxed to stressed activity in the nonparametric statistical tests and Cohen's D. The VLF power is the percentage of power in the Very Low 289 Frequency band $(0.003 \le f \le 0.04 \text{ Hz})$ of the frequency domain transformation (Fast Fourier 290 Transform) of the time domain RR (interval between heart beats) interval data ($p = 0.0098$, d = 0.56, a medium size effect). The degree of heart rate variability (HRV) in terms of VLF power 292 is expected to reduce as activity intensity increases, an effect observed in our data. The 293 minimum heart rate was lower during relaxed activity ($p = 0.016$, $d = 0.24$, a small size effect). Sample entropy, which indicates randomness of the data (lack of regularity in heart 295 rate), was higher for the stressed activity ($p=0.014$, $d=0.2$, a small size effect). Finally, the mean RR was significantly higher during the relaxed activity, indicating greater heart rate 297 variability and hence less stress ($p = 0.025$, $d = 0.2$, a small size effect). Results of subjective feedback assessments (sleepiness, thermal preference) and pupil dilation data are provided in the Supporting Information. Mean and median pupil size were significantly larger during stressed than relaxed activity, shown in Figure S4 of the Supporting Information, another objective indicator associated with stress.⁴⁹ Collectively, these objective indicators demonstrate that cognitive testing vs. nature documentary was effective in causing a relative increase in participants' stress.

304 The end-tidal CO_2 (etCO₂) and respiration rate data did not show significant difference 305 between relaxed and stressed activity periods $(p > 0.7)$, determined using a t-test since distributions were normal. This finding was somewhat unexpected given that emission rates of $307 \quad \text{CO}_2$ calculated from chamber air during occupied periods (Section 4.3) are elevated during 308 stressed compared to relaxed activity. Importantly, $etCO₂$ was measured post-testing, over a

period of ~10 minutes following completion of relaxed or stressed activity while bioeffluent

CO² emission rates are calculated with data collected during the 30-min of relaxed or stressed

311 activity. No significant changes in etCO₂ implies body CO_2 levels were similar at the end of

312 the two periods. As respiration rates are also similar, in conjunction with etCO_2 values, this

313 indicates that the body is able to effectively ventilate the additional $CO₂$ being generated

during the test period, possibly through breathing in greater volumes of air.

4.2 Human emissions of CO² and isoprene

Representative experimental results from one of the four days of controlled testing of

human chemical emissions are shown in Figure 3. An abbreviated experimental protocol is

shown: measurement of levels of chemical emissions occurred during a baseline (unoccupied)

period and two subsequent occupied periods. Also shown in Figure 3 is example TD-GC-MS

data for isoprene, which is time-integrated. The x-direction bars on the plot indicate the period

of time over which the sample was taken while the y-direction error bars show the range

across duplicate samples taken in chamber air.

323
324 Figure 3. Representative experimental results for chamber dry-bulb air temperature, relative 325 humidity, isoprene levels during baseline and occupied periods and time-series measurements 326 of carbon dioxide.

 327 The effect of occupancy on $CO₂$ levels and the flushing of the chamber in between the 328 two occupied periods is apparent in Figure 3, the flushing due to a temporarily placed fan that 329 evacuated air from the chamber, replacing with ambient laboratory air and reducing to near-330 background levels of $CO₂$ prior to the initiation of the second test activity. The test chamber 331 CO² level was monitored and observable in real-time, and was used to ensure the second test 332 activity of each day occurred with initial $CO₂$ conditions similar to that of the first test. The 333 short duration for flushing the chamber between occupied activities is one reason for limiting 334 analysis of chemical emissions to $CO₂$ and isoprene, both compounds that are predominantly 335 present in the gas-phase. There existed no aqueous water present in the chamber and surfaces 336 were impermeable, hard surfaces. $CO₂$ sorption to surfaces indoors is typically taken as

347 We calculated source strengths during baseline, relaxed and stressed activities for CO²

348 and isoprene. Results of the source strength calculation are shown in Table 1 for isoprene and

350 **Table 1.** Chamber conditions and magnitudes of carbon dioxide and isoprene emissions 351 across baseline, relaxed, and stressed activities.

 *CO² levels were measured in 1-min interval in the inlet and room, averages over the 30-min test periods are reported here for brevity. See Figure S5 for full time-series analysis of CO² data. $\frac{4}{355}$ # baseline CO_2 and isoprene emission rates are presented on a per-person normalized basis, though no humans were present in the chamber during baseline conditions, to enable comparisons to the occupied (relaxed and stressed activities). All per-person normalizations were made by dividing the total emission rate into the chamber by five, four participants plus the experimenter. % isoprene levels were corrected for field blanks in calculations of isoprene source strength 361 (field blanks were 0.27, 0.27, 0.27, and 0.45 μ g/m³ on days, 1, 2, 3 and 4, respectively) Emission rates calculated from mass-balance models during accumulation of $CO₂$ in the test chamber (28.7 g/h/p averaged across all participants and conditions) are in agreement with prior estimates of per-person $CO₂$ generation rates. Values reported here are consistent with estimates for adults in residences and bedrooms, and slightly lower than values typically 367 assumed for office environments.^{1,18} Per-person $CO₂$ emission rates reported here are similar 368 to those reported by Stönner et al.²⁸ for adults (30 g/h/p) and Wang³⁵ for students (27 g/h/p) during lecture, 38 g/h/p during exam). Our results are somewhat higher than the average 370 reported by Tang et al. ² in an occupied college classroom (21 g/h/p). Our per-person 371 emission rate for isoprene is 135 µg/h/p across all occupied conditions (132 µg/h/p if the aforementioned baseline emission is accounted for), are in general agreement with prior studies. Three independent prior studies reporting per-person isoprene emission rates for 374 adults are 105, 162, and 166 μ g/h/p.^{2,27,28} Our estimate for per-person isoprene emission rate is near the average of these prior studies. *4.3 Impact of cognitive tasks on isoprene and CO² emission rate*

-
- Full results across the four days of repeated testing are shown in Figure 4 for carbon dioxide and isoprene. Per-person CO² emission rates were greater for stressed than relaxed 380 conditions $(30.3 \pm 2.1 \text{ vs. } 27.0 \pm 1.7 \text{ g/h/p}, p = 0.0044 \text{, mean } \pm \text{ standard deviation})$ and

381 isoprene emission rates were also elevated under stressed vs. relaxed conditions $(154 \pm 25$

 Figure 4. Per-person emission rates of carbon dioxide and isoprene. The horizontal bar is the average across all days of testing, each dot shows the measurement for a particular day. Values of emission rates shown for occupied condition emissions (i.e., relaxed and stressed activities) do not subtract unoccupied (baseline) estimates of emissions.

 Shown in Figure 4 is that carbon dioxide emission rates are consistently higher (12%) during stressed activity than relaxed activity for all repeated trials of the study; regressions 391 used to estimate emission rates from dynamic $CO₂$ data each day of testing are shown in Figure S5 of the Supporting Information. Because the study was counterbalanced, consistently 393 elevated CO_2 emissions imply that it is the stressed activity that caused the elevated CO_2 emission, rather than the order of tests conducted. Inspection of isoprene emission rates also reveals elevated emission rate under stressed activity; isoprene emissions were 33% higher 396 than the relaxed activity. The finding of greater relative increases in isoprene vs. $CO₂$ emission rate is interesting given that demand-controlled ventilation systems are typically 398 based on $CO₂$ levels; results here indicate isoprene may be a useful additional parameter for designing building ventilation systems with sensitivity to cognitive tasks.

 As presented in Section 4.1, we measured objective indicators of occupant stress, including biomarkers of salivary alpha amylase and metrics associated with participant's heart rate variability and pupil dilation. We also measured chamber temperature and relative humidity conditions. To explore the utility of these objective biomarkers and environmental conditions as possible predictors of the observed elevation in bioeffluent emission rates, we 405 performed regression to observe relationships to variance and change in $CO₂$ and isoprene emission rates. Full results of this analysis are shown in Figure S6 of Supporting Information for biomarkers observed to be statistically significantly different between stressed and relaxed activity and temperature and RH. Note that since bioeffluent emission rates are the average per-person emission for individuals present in the chamber, we regress these emission rates against biomarkers averaged for those same individuals. The sAA data here includes that of the experimenter who was inside the chamber, was not subject to the same research protocol, 412 but nevertheless contributed to $CO₂$ and isoprene emissions. However, since the experimenter did not wear an HRV monitor, the HRV related markers do not have his information. We found that of biomarkers measured, only subject salivary alpha amylase (Figure 5) 415 is significantly related to isoprene emission rates (p -value = 0.02), and that a substantial 416 portion of the variance in observed isoprene emission rates is explained ($\mathbb{R}^2 = 0.60$). This result is interesting considering a prior study identifies isoprene as a marker associated with 418 – emotional stress arising from moviegoers' interaction with audiovisual stimuli during films.³⁶ 419 Stönner et al.³⁷ suggest that isoprene emission and variability in isoprene level could be useful as an objective indicator for recommending age-appropriateness of different films. Isoprene is 421 thought to be generated during cholesterolgenesis⁵² and some studies suggest measurement of breath isoprene levels may be useful as a non-invasive method for assessing blood cholesterol

423 levels.⁵³ Williams et al.³⁶ suggest isoprene production is related to cortisol production via the cholesterol production pathway; this link appears based on their empirical findings, as no biological pathway is mentioned. If isoprene and cortisol production are related, this may explain the observed correlation here between isoprene emission and salivary alpha amylase, since studies show responses of both salivary alpha amylase and cortisol follow stressful 428 events with salivary alpha amylase levels responding more rapidly than cortisol levels.⁵⁴ However, it should be cautioned that linking the production of an endogenously generated compound like isoprene to external stimuli is challenging, as per-person differences in isoprene production as measured by human breath may result from interactions between the 432 metabolic pathways with the circulatory (heart rate) and pulmonary systems (breathing rate).⁵³

433
434 Figure 5. Regression of isoprene and CO₂ emission rates vs. salivary alpha amylase and chamber air temperature. Shown above each plot are the r-squared and *p*-value associated with the linear regression slope. Curves around each linear regression are the 95% confidence 437 interval. Note that the target chamber temperature was constant at $26 °C$; the above plot investigates whether incidental variation in temperature is related to observed changes in in 439 emission rates of isoprene and CO₂.

 relationship is not significant. A linear regression of emission rate vs. relative humidity also resulted in no significant correlation (Figure S6 of Supporting Information).

470 In this study, we limit our analysis to isoprene and $CO₂$ as robust quantification of other VOCs is limited by the calibration standard and TD-GC-MS method applied to this chamber study. One further limitation is that our dynamic analysis of isoprene emission rates was constrained by time-integrated sampling in the chamber and inlet air. We believe the method for estimating per-person isoprene emission rates developed is robust; empty chamber isoprene emission rates were near zero while occupied chamber emission rates are in close agreement with prior estimates. However, there exists uncertainty associated with the initial conditions required for the solution, that is, the level of isoprene was assumed to be that of the baseline conditions at the start of an occupied chamber experiment. Sensitivity analyses indicate the solution is relatively insensitive to this input; model runs show a 20% change to 480 the initial condition in the chamber results in a \sim 5% change in the calculated isoprene emission rate. As discussed, surface adsorption/desorption of isoprene may impact the second occupied test of each day, though we believe the effect is likely small based on the high volatility of isoprene, our observation of near-zero isoprene emission rates in the unoccupied (baseline) test chamber, and counterbalanced design.

 Subsequent chamber studies should employ additional analytical instrumentation, e.g., real-time chemical ionization mass spectrometry, to further elucidate the chemical complexity of human chemical emissions impacted by cognitive task. These instruments would greatly expand the classes of compounds that can be measured and improve robustness of source strength calculations with dynamic measurements. Future studies should also be conducted on larger sample populations, with different cognitive tasks, and examine potential for

 compounding interactions with other indoor environmental quality variables, e.g., varying temperature and RH condition.

Supporting Information

The Supporting Information contains additional information concerning the test bed facility,

experimental conditions, details of cognitive tests used for the "stressed" activity,

characteristics of participants, and additional details of indicators of participant stress and

feedback. Further information is provided concerning the measurement of isoprene and

resulting calculation of isoprene emission rates. Additional data is reported from the

499 subjective feedback and pupillometry, analysis of $CO₂$ regressions to determine $CO₂$ emission

500 rates, and full results of regression of $CO₂$ and isoprene emission rates with stress indicators

and chamber temperature and RH conditions.

Author Contributions

 ETG led study design, analysis, and writing, AM contributed to study design, analysis and writing, JL contributed to study design, analysis and data visualization, SS contributed to study design, writing, and editing, AL contributed to data analysis and editing. All authors have given approval to the final version of the manuscript.

Acknowledgements

This research was funded by the Republic of Singapore's National Research Foundation

through the SinBerBEST program. The research was carried out within the SinBerBEST

Testbed [\(http://sinberbest.berkeley.edu/content-page/testbed-facilities\)](http://sinberbest.berkeley.edu/content-page/testbed-facilities). We thank the

anonymous reviewers for their thoughtful and thorough comments that improved the paper.

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