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Impact of cognitive tasks on CO₂ and isoprene 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 CO_2 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
- activity). Measured biomarkers indicated higher stress levels were achieved during the
- 18 stressed activity. Per-person CO_2 emission rates were greater for stressed than relaxed activity
- 19 (30.3 ± 2.1 vs. 27.0 ± 1.7 g/h/p, p = 0.0044, mean ± standard deviation). Isoprene emission
- rates were also elevated under stressed vs. relaxed activity $(154 \pm 25 \,\mu g/h/p \text{ vs. } 116 \pm 20 \,\mu g/h/p, p = 0.041)$. Chamber temperature was held constant at $26.2 \pm 0.49 \,\circ\text{C}$; incidental
- $\mu g/\mu p, p = 0.041$). Chamber temperature was held constant at 20.2 \pm 0.49 °C, incidentative variation in temperature did not explain variance in emission rates. Isoprene emission rates
- increased linearly with salivary-alpha amylase levels ($r^2 = 0.6$, p = 0.02). These results imply
- the possibility of considering cognitive tasks when determining building ventilation rates.

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

- 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
- 31 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
- indicated by proxy measurement of carbon dioxide (CO_2) , are also a driver of the need to
- 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
Exposure to human chemical emissions may also impact human cognition.^{8,9} Studies have
implicated exposure to pure elevated CO₂ as impairing human cognition,^{10,11} though there also
exist studies that show no effect of CO₂ itself on cognition.^{12,13} However, it is consistently
shown that reduced outdoor air ventilation, leading to higher indoor concentration of human
chemical emissions, is responsible for observed decrements in cognition or environment
perception.¹⁴⁻¹⁶

A variety of factors beyond ventilation rate and occupant density impact the level of 42 43 human chemical emissions in a given indoor space. There are methods for predicting the CO₂ generation rate per-occupant,^{17,18} and they consider body surface area, body composition, 44 metabolic rate (related to type of activity), air temperature, and a respiratory quotient, which is 45 largely a function of diet.¹⁹ There are many studies in the literature quantifying type and 46 quantity of emissions of endogenous and exogenous volatile organic compounds,^{20–23} often in 47 pursuit of understanding the human "volatilome" for diagnosis of disease.²⁴ Enabled by 48 analytical methods that can measure VOCs with high time resolution, studies have estimated 49 average, per-person emission rates of carbon dioxide and VOCs in a variety of field settings 50 including classrooms,^{2,25,26} a museum,²⁷ and a movie theater²⁸; these studies have substantially 51 expanded our knowledge of the magnitudes and type of human chemical emissions. However, 52 53 comparatively fewer studies investigate factors in healthy humans that may influence human chemical emissions, including traits like age, sex, or smoking status,^{29–31} and their 54 biochemical mechanisms.³² Recent studies, conducted in controlled chambers, measured 55 56 human emissions under varying "human factors" including type of clothing worn and age, as well as environmental factors, including temperature.^{33,34} 57

The impact of psychological factors on human chemical emissions is not well-studied 58 or understood. In a 1975 field study of human bioeffluent emissions, Wang et al.³⁵ measured 59 volatile emissions from students in a University classroom used for lectures and examinations. 60 They found twelve organic compounds were elevated during lecture periods, and estimate per-61 person production rates. In that study, the researchers identified and separated examination 62 63 periods as a condition where occupants experienced increased stress relative to lecture. They reported a 43% increase in CO_2 emission rate when comparing examination to lecture 64 65 conditions; the authors suggest that this was due to elevated metabolism. More recently, Williams et al.³⁶, investigated the role of audiovisual stimuli to influence human chemical 66 67 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 68 theater. By annotating films with scene content labels, the researchers ascertained whether a 69 70 scene label could predict variation of the measured human chemical emissions. Carbon 71 dioxide and isoprene exhibited high correlations with scene labels, indicating emissions of these compounds may be associated with biochemical pathways induced by emotional 72 responses to the movie scene. Isoprene emissions and variability have also been studied as a 73 potential marker for age-appropriateness of film content,³⁷ and may serve as an indicator for 74 several metabolic processes in the human body, including lipid metabolism and physiological 75 state.38 76

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 81 of the occupants. Conversely, monitoring levels of human chemical emissions in buildings 82 could be used to gauge occupant cognitive engagement. To date, studies advancing our 83 knowledge of factors influencing human chemical emissions have generally occurred in 84 uncontrolled field environments (movie theaters, classrooms, museums). We use a state-of-85 86 the-art controlled test chamber to engage sixteen subjects in an across subjects, counterbalanced study, to explore how cognitive tasks may influence human chemical 87 88 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. 89

90 **3.** Methods and materials

91 *3.1 Test chamber and subjects*

We conducted studies in a room-scale test chamber to measure the influence of 92 cognitive tasks on CO₂ and isoprene emitted by humans. The test chamber is located within 93 the Singapore-Berkeley Building Efficiency for Sustainability in the Tropics facility (Figure 94 95 S1, Supporting Information). The test bed consists of four climatic chambers, each $5.6 \times 4.3 \times 2.7$ m. Our tests were conducted in one of the four chambers mocked up as a typical 96 97 office environment with desks and computer workstations. Four participants and one 98 researcher were present during test periods, shown in Figure 1 (a) and (b). An experimental 99 timeline, post-occupancy by test subjects, is shown in Figure 1 (c). The test chamber was 100 served by a single-pass, mixing ventilation forced air heating, ventilation, and air-conditioning 101 (HVAC) system that includes supply air cleaning, with MERV8 and pleated activated carbon 102 (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 103 104 ceiling fan with transparent blades (diameter = 1.47 m, Artemis Minka-Aire F803L-TL model,

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



108

Figure 1. Overview of experimental set-up (a) and (b) and time-line of experiments postoccupancy (c). SF = subjective feedback, ETC = eye tracker calibration, SF-C = subjective feedback & capnometer, TCS = thermal comfort monitoring station, CO_2 = Wall mounted CO₂ 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

118 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 119 120 in a 30-min series of cognitive tasks. We selected five cognitive tasks covering the executive functions: working memory, inhibition, and cognitive flexibility/task switching.³⁹ The tasks 121 were administered through the PEBL platform.⁴⁰ Prior studies suggest that audiovisual stimuli 122 can alter stress state, observable by biomarkers like cortisol and salivary alpha amylase,^{41,42} 123 with viewing of stressful or soothing videos of ~ 30 min in length.^{42,43} We include further details 124 of cognitive tests used in the Supporting Information. In total, we tested sixteen participants in 125 groups of four, with the order of exposure to the relaxed and stressed activities alternating over 126 127 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 128 of California Berkeley Ethics Committee (Protocol #IRB2019-07-12403). Participants were 129 130 briefed on the study in a one-hour familiarization session conducted for each participant prior to the experiments. 131

We measured objective and subjective indicators of occupant stress and perceptions of 132 the test-chamber's indoor environment. Participants arrived 15 minutes prior to the start of the 133 session and were outfitted with a heart rate sensor (Polar H10, Polar, Finland). During the 134 135 session, each participant sat at a workstation with a laptop. Each laptop had an eye tracker 136 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 137 138 Corporation, Japan) and end tidal CO₂ (etCO₂) level (CO₂ monitor, OLG-3800, Nihon Kohden, Japan). We also collected subjective feedback concerning indoor environmental 139 140 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

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

143 *3.3 Human chemical emission measurement*

Human chemical emissions evaluated in this study include carbon dioxide and 144 isoprene. Human chemical emissions were measured during a 30-min baseline period, 145 occurring prior to the arrival of occupants, and during the two 30-min periods during 146 147 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 30 149 150 ppm and 3%). Time-integrated measurements of volatile organic compounds were made 151 during baseline and occupied periods via sampling with sorbent tubes packed with 100 mg of Tenax TA and 180 mg of Carbograph 1, based on a method developed previously,⁴⁴ modified 152 153 for target compounds, sampling times, and sampling flowrate (~130 mL/min, total sample 154 volume = 4 L). One portable sampling pump (PCXR4, SKC, USA) drew sample air through 155 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 ~2 m of 0.635" OD tubing placed upstream of the sorbent cartridge. Bulk room air samples were made in 157 158 duplicate using a flow splitter. Sorbent cartridges were analyzed with thermal desorption – gas 159 chromatography – mass spectrometry (TD-GC-MS, TurboMatrix 650, PerkinElmer, USA connected to a gas chromatograph, Model 7890 A, Agilent Technologies, USA). We 160 161 quantified VOCs by generating calibration curves from a calibration mixture that included 162 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 163 164 difference (RPD) across all duplicate isoprene concentrations measured was 10.9% while the

165 RPD for acetone duplicates was 66.8%. Since acetone is emitted by humans with high source 166 strength, we suspect the method is not suitable for robust acetone quantification at the levels 167 present in the chamber. Further details of the sampling and analysis method are provided in 168 the Supporting Information and in Figure S2. To return chamber bioeffluent levels to near-169 background levels, during the break between the two activities the room was flushed with a 170 fan (flowrate = 62 m³/min, IBF300, Aman, India) providing ~10 turnovers of air during the 171 15-min break.

172 Chamber temperature and relative humidity were measured continuously in 1-min 173 intervals (ThermCondSys 5500, Sensor, Poland) and ozone levels were monitored (Model 174 205, 2BTech, USA) once per day, prior to the arrival of occupants, to ensure near-zero ozone 175 levels; ozone was consistently <2 ppb, the stated uncertainty of the instrument. Light levels at 176 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 177 178 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 179 180 tests conducted following the conclusion of the experiments for the day (excepting day 1) (see 181 Table S1). The chamber is well-mixed, confirmed via measurements of temperature and airflow gradients, as well as high r^2 (>0.99) when applying a well-mixed model to CO₂ tracer 182 183 decay periods. The sorbent tubes and sampling pumps were located in a chamber adjacent to 184 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). 185

186 *3.4 Data analysis*

187 *3.4.1 Calculation of CO*₂ *emission rates*

Time-series measurements of CO_2 were used to estimate emission rates of CO_2 for the baseline and 30-minute periods during which participants were engaged in either relaxed or stressed activity. We used a non-linear curve-fit applied to the analytical solution of a massbalance equation that included air-exchange and CO_2 emissions from occupants, shown in equation 1:

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

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

A best-fit value of S_{co2} over the 30-min period was determined using the "fit" function in MATLAB R2019a, with the fit type specified as eq. 1. The chamber was occupied by five persons during all occupied test conditions; we report CO₂ source strengths on a per-person basis by dividing the best-fit S_{co2} from eq. 1 by five. Note that one of the occupants was always the same researcher, present in the chamber as per IRB protocol requirement. He maintained a similar diet, exercise level, and sleep schedule during the week of testing, and endeavored to maintain similar during-test activity level across all tests.

206 *3.4.2 Calculation of isoprene emission rates*

Isoprene was measured over three 30-min periods: baseline (unoccupied), stressed and relaxed activity (both occupied by five persons). Measured isoprene levels are time-integrated 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:

$$S_{isoprene,baseline} = \lambda (C_{i,chamber} - C_{i,o}) V$$
⁽²⁾

where $S_{isoprene,baseline}$ is the source or sink strength of isoprene (µg h⁻¹) under baseline conditions, λ is the air-exchange rate due to ventilation of the chamber (h⁻¹), C_i is the concentration of isoprene in chamber air (µg m⁻³), $C_{i,o}$ is the concentration of isoprene in supply air (µg 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 219 220 strength using a mass-balance model as in equation 1, written for isoprene. Because our 221 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 222 stressed activity period to determine the area under the concentration-time curve. Note that the 223 224 relaxed or stressed condition began 10 min following occupancy (Figure 1c). This area 225 calculation was established as the objective function, and the source strength of isoprene over 226 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: 227

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

curve (1 min), $C_{i,measured}$ is the time-integrated measurement of isoprene in the chamber, taken as the average of duplicate measurements of isoprene and corrected for field blank (µg m⁻³), and $\Delta t_{measured}$ is the time duration of the measured isoprene level in the chamber (30min).

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.

238 *3.4.3 Statistical tests*

239 As shown in Table S1 of the Supporting Information, we conducted tests over four days where CO₂ and isoprene source or sink strengths were measured for baseline, relaxed, 240 241 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 243 source or sink strength in the chamber, we used analysis of variance (ANOVA) across three groups (baseline, relaxed, stressed), similar to a method described previously.⁴⁵ Briefly, if the 244 three-group comparison resulted in $\alpha < 0.05$, we then made direct comparisons across groups 245 246 (i.e., baseline vs. relaxed, baseline vs. stressed, and relaxed vs. stressed) using Tukey's honest 247 significant difference test (Tukey test). We included the baseline (unoccupied) conditions in ANOVA to evaluate whether source or sink strengths during occupied conditions were 248 significantly different from unoccupied conditions. All statistical analyses were performed 249 using MATLAB 2019a. 250

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,
- 255 Benjamini–Hochberg procedure (false discovery rate taken as 10%) was used. All
- comparisons were paired and two tailed.
- 257 Data and analysis scripts/functions are publicly available and can be downloaded from
- the Dryad repository: https://doi.org/10.5061/dryad.gb5mkkwmk.
- 259 4. Results and Discussion
- 260 *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
- 267 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.





275 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 277 278 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 279 280 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 281 this effect to be small based on the diurnal profiles presented by Nater et al.⁴⁶. We analyzed 282 283 51 parameters of heart rate variability; as noted from previous studies,⁴⁷ not all parameters are expected to yield a significant difference when stress is imposed on people. Note that one 284 participant's data was lost on day 3; subsequent analysis used data from fifteen participants. In 285

286 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 287 statistical tests and Cohen's D. The VLF power is the percentage of power in the Very Low 288 Frequency band $(0.003 \le f < 0.04 \text{ Hz})$ of the frequency domain transformation (Fast Fourier 289 290 Transform) of the time domain RR (interval between heart beats) interval data (p = 0.0098, d =291 0.56, a medium size effect). The degree of heart rate variability (HRV) in terms of VLF power is expected to reduce as activity intensity increases,⁴⁸ an effect observed in our data. The 292 minimum heart rate was lower during relaxed activity (p = 0.016, d = 0.24, a small size 293 294 effect). Sample entropy, which indicates randomness of the data (lack of regularity in heart rate), was higher for the stressed activity (p=0.014, d=0.2, a small size effect). Finally, the 295 296 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 298 299 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 300 objective indicator associated with stress.⁴⁹ Collectively, these objective indicators 301 302 demonstrate that cognitive testing vs. nature documentary was effective in causing a relative 303 increase in participants' stress.

The end-tidal CO₂ (etCO₂) and respiration rate data did not show significant difference 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 CO₂ calculated from chamber air during occupied periods (Section 4.3) are elevated during 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

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

activity. No significant changes in $etCO_2$ implies body CO_2 levels were similar at the end of

the two periods. As respiration rates are also similar, in conjunction with etCO₂ values, this

313 indicates that the body is able to effectively ventilate the additional CO_2 being generated

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

315 *4.2 Human emissions of CO*₂ 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

318 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

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

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

322 across duplicate samples taken in chamber air.



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

323

327 The effect of occupancy on CO_2 levels and the flushing of the chamber in between the two occupied periods is apparent in Figure 3, the flushing due to a temporarily placed fan that 328 329 evacuated air from the chamber, replacing with ambient laboratory air and reducing to nearbackground levels of CO_2 prior to the initiation of the second test activity. The test chamber 330 331 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 CO_2 conditions similar to that of the first test. The short duration for flushing the chamber between occupied activities is one reason for limiting 333 analysis of chemical emissions to CO_2 and isoprene, both compounds that are predominantly 334 present in the gas-phase. There existed no aqueous water present in the chamber and surfaces 335 were impermeable, hard surfaces. CO_2 sorption to surfaces indoors is typically taken as 336

337	negligible. Isoprene is considered a very volatile organic compound, or VVOC, with vapor
338	pressure 7.2×10^4 Pa. ⁵⁰ A recently published model of partitioning of organics indoors shows
339	isoprene is expected to be predominantly present in the gas-phase across a range of indoor
340	surface reservoir conditions, though experimentally determined response times indicate
341	dynamic partitioning does occur. ⁵¹ We expect that error introduced for isoprene from surface
342	partitioning is likely small, based on calculations of emission rates into the chamber under
343	unoccupied conditions vs. occupied conditions (13 μ g/h vs. 675 μ g/h). Furthermore,
344	uncertainty due to the impact of isoprene surface interactions on comparisons of emission
345	magnitudes across stressed and relaxed activities is addressed, at least partially, in the
346	counterbalanced design.

We calculated source strengths during baseline, relaxed and stressed activities for CO₂
and isoprene. Results of the source strength calculation are shown in Table 1 for isoprene and

349 CO ₂	, which met	the three-group	o comparison	ANOVA	criteria	of α <	(0.05.
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Day		Temp	RH	O₃	CO ₂ inlet*	CO₂ chamber*	lsoprene inlet [%]	lsop charr	rene 1ber [%]	CO ₂ emission rate	Isoprene emission rate
		۰C	%	ppb	ppm	ppm	µg/m³	µg/m³		g/h/p	µg/h/p
								rep1	rep2		
	Baseline [#]	26.9	38	<2	415	420	2.1	1.5	2.1	0.32	8.03
1	Relaxed	26.9	43		441	784	2.4	4.9	5.4	27.8	144.2
	Stressed	27.1	48		467	768	2.8	5.6	5.7	30.5	154.9
	Baseline [#]	25.6	41	<2	411	435	1.8	1.9	1.9	0.65	-2.72
2	Stressed	26.2	46		406	774	1.9	4.8	3.9	32.3	119.3
	Relaxed	26.5	47		441	722	1.9	4.5	4.0	27.7	115.6
	Baseline [#]	25.5	41	<2	403	415	1.6	1.7	1.4	0.58	2.84
3	Relaxed	26.1	45		421	733	2.0	3.9	3.9	25.4	102.5
	Stressed	26.4	48		435	721	2.6	5.7	6.0	28.9	176.4
4	Baseline [#]	25.5	40	<2	403	418	2.2	2.5	2.2	0.69	2.75
	Stressed	26.2	45		420	780	2.6	5.8	6.1	29.7	165.9
	Relaxed	26.5	47		430	712	2.1	4.1	4.5	27.3	100.0

Table 1. Chamber conditions and magnitudes of carbon dioxide and isoprene emissionsacross baseline, relaxed, and stressed activities.

 $*CO_2$ levels were measured in 1-min interval in the inlet and room, averages over the 30-min 352 353 test periods are reported here for brevity. See Figure S5 for full time-series analysis of CO₂ 354 data. 355 # baseline CO₂ 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 356 comparisons to the occupied (relaxed and stressed activities). All per-person normalizations 357 were made by dividing the total emission rate into the chamber by five, four participants plus 358 359 the experimenter. % isoprene levels were corrected for field blanks in calculations of isoprene source strength 360 (field blanks were 0.27, 0.27, 0.27, and 0.45 μ g/m³ on days, 1, 2, 3 and 4, respectively) 361 362 363 Emission rates calculated from mass-balance models during accumulation of CO_2 in the test chamber (28.7 g/h/p averaged across all participants and conditions) are in agreement 364 365 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 366 assumed for office environments.^{1,18} Per-person CO₂ emission rates reported here are similar 367 to those reported by Stönner et al.²⁸ for adults (30 g/h/p) and Wang³⁵ for students (27 g/h/p 368 during lecture, 38 g/h/p during exam). Our results are somewhat higher than the average 369 reported by Tang et al.² in an occupied college classroom (21 g/h/p). Our per-person 370 emission rate for isoprene is 135 μ g/h/p across all occupied conditions (132 μ g/h/p if the 371 aforementioned baseline emission is accounted for), are in general agreement with prior 372 373 studies. Three independent prior studies reporting per-person isoprene emission rates for adults are 105, 162, and 166 μ g/h/p.^{2,27,28} Our estimate for per-person isoprene emission rate 374 375 is near the average of these prior studies. 376 4.3 Impact of cognitive tasks on isoprene and CO₂ emission rate 377

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 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

isoprene emission rates were also elevated under stressed vs. relaxed conditions (154 ± 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.

389 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 390 391 used to estimate emission rates from dynamic CO₂ data each day of testing are shown in 392 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 394 395 reveals elevated emission rate under stressed activity; isoprene emissions were 33% higher than the relaxed activity. The finding of greater relative increases in isoprene vs. CO₂ 396

- 397 emission rate is interesting given that demand-controlled ventilation systems are typically
- based on CO₂ levels; results here indicate isoprene may be a useful additional parameter for
- designing building ventilation systems with sensitivity to cognitive tasks.

400 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 401 rate variability and pupil dilation. We also measured chamber temperature and relative 402 humidity conditions. To explore the utility of these objective biomarkers and environmental 403 404 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 406 407 for biomarkers observed to be statistically significantly different between stressed and relaxed 408 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 409 against biomarkers averaged for those same individuals. The sAA data here includes that of 410 411 the experimenter who was inside the chamber, was not subject to the same research protocol, but nevertheless contributed to CO₂ and isoprene emissions. However, since the experimenter 412 413 did not wear an HRV monitor, the HRV related markers do not have his information. 414 We found that of biomarkers measured, only subject salivary alpha amylase (Figure 5) is significantly related to isoprene emission rates (p-value = 0.02), and that a substantial 415 portion of the variance in observed isoprene emission rates is explained ($R^2 = 0.60$). This 416 result is interesting considering a prior study identifies isoprene as a marker associated with 417 emotional stress arising from moviegoers' interaction with audiovisual stimuli during films.³⁶ 418 Stönner et al.³⁷ suggest that isoprene emission and variability in isoprene level could be useful 419 as an objective indicator for recommending age-appropriateness of different films. Isoprene is 420 thought to be generated during cholesterolgenesis⁵² and some studies suggest measurement of 421 422 breath isoprene levels may be useful as a non-invasive method for assessing blood cholesterol

levels.⁵³ Williams et al.³⁶ suggest isoprene production is related to cortisol production via the 423 cholesterol production pathway; this link appears based on their empirical findings, as no 424 425 biological pathway is mentioned. If isoprene and cortisol production are related, this may 426 explain the observed correlation here between isoprene emission and salivary alpha amylase, 427 since studies show responses of both salivary alpha amylase and cortisol follow stressful events with salivary alpha amylase levels responding more rapidly than cortisol levels.⁵⁴ 428 429 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 430 431 isoprene production as measured by human breath may result from interactions between the metabolic pathways with the circulatory (heart rate) and pulmonary systems (breathing rate).⁵³ 432



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 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 emission rates of isoprene and CO₂.



446	Figure S6 in the Supporting Information, we do observe that the relation between isoprene
447	emission rates and three of the four HRV related markers are in the expected directions, that
448	is, human chemical emission of isoprene increases with increases in stress markers, though
449	not statistically significantly.
450	Prior studies have explored the relationship between cognitive tests and other factors
451	on human VOC emissions. Santos et al. ⁵⁵ , with a study design that employed a cognitive test
452	to elicit stress in a cohort of 14 individuals, identified possible VOCs that are indicative of
453	stress. Their study identified benzaldehyde, ethyl acetate, and 2-propanol as indicators of
454	stress, compounds we found in the chamber at measurable levels in TD-GC-MS samples.
455	Santos et al. ⁵⁵ employed a prototype sampling device and analysis via GC-IMS. The GC-IMS
456	method is noted as subject to uncertainty in identification of compounds. We do not here
457	speculate concerning the possible relationship of these compounds to cognitive tasks, since
458	these compounds were not present in our calibration mixture.
459	A recent study explored the relationship between temperature and human chemical
460	emissions, finding increases in ammonia emissions from humans with increasing
461	temperature. ³⁴ Our study design aimed to hold temperature constant; temperature averaged
462	26.2 ± 0.49 °C (mean ± standard deviation). Given the known association between air
463	temperature and human CO_2 emissions ¹⁸ and recent observation of temperature dependence
464	on ammonia emissions, Figure 5 includes a regression of CO_2 and isoprene emission rates as a
465	function of chamber air temperature, plotted as the natural log of emission rate vs. the inverse
466	of temperature. While slopes are in the expected direction (i.e., a negative slope, implying
467	higher temperature yields higher emission rate), the regression statistics indicate the

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 CO_2 as robust quantification of other VOCs is limited by the calibration standard and TD-GC-MS method applied to this 471 472 chamber study. One further limitation is that our dynamic analysis of isoprene emission rates 473 was constrained by time-integrated sampling in the chamber and inlet air. We believe the 474 method for estimating per-person isoprene emission rates developed is robust; empty chamber 475 isoprene emission rates were near zero while occupied chamber emission rates are in close 476 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 477 478 baseline conditions at the start of an occupied chamber experiment. Sensitivity analyses 479 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 481 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 482 volatility of isoprene, our observation of near-zero isoprene emission rates in the unoccupied 483 484 (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

491 compounding interactions with other indoor environmental quality variables, e.g., varying492 temperature and RH condition.

493 Supporting Information

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

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

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

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

498 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

rates, and full results of regression of CO_2 and isoprene emission rates with stress indicators

and chamber temperature and RH conditions.

502 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.

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