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The impact of plug-in fragrance diffusers on residential indoor VOC concentrations†

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Plug-in fragrance diffusers are one of myriad volatile organic compound-containing consumer products that are commonly found in homes. The perturbing effects of using a commercial diffuser indoors were evaluated using a study group of 60 homes in Ashford, UK. Air samples were taken over 3 day periods with the diffuser switched on and in a parallel set of control homes where it was off. At least four measurements were taken in each home using vacuum-release into 6 L silica-coated canisters and with >40 VOCs quantified using gas chromatography with FID and MS (GC-FID-QMS). Occupants self-reported their use of other VOC-containing products. The variability between homes was very high with the 72 hour sum of all measured VOCs ranging between 30 and >5000 $\mu\text{g m}^{-3}$, dominated by *n*-i-butane, propane, and ethanol. For those homes in the lowest quartile of air exchange rate (identified using CO₂ and TVOC sensors as proxies) the use of a diffuser led to a statistically significant increase (*p*-value < 0.02) in the summed concentration of detectable fragrance VOCs and some individual species, e.g. alpha pinene rising from a median of 9 $\mu\text{g m}^{-3}$ to 15 $\mu\text{g m}^{-3}$ (*p*-value < 0.02). The observed increments were broadly in line with model-calculated estimates based on fragrance weight loss, room sizes and air exchange rates.

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Environmental significance

The effects of using of a single VOC-emitting domestic product – a plug-in fragrance diffuser – on changing indoor concentrations of VOCs were evaluated in 60 homes. Its use led to detectable increases in α -pinene of a few $\mu\text{g m}^{-3}$ in homes with the lowest air exchange rates (inferred from indoor CO₂). A model was developed that estimated indoor increments in VOC concentrations for a given product emission rate, room size and air changes per hour.

1. Introduction

The use of room fragrance products is an acknowledged source of emissions of a range of different volatile organic compounds (VOCs) indoors. Quantifying the emission rate characteristics and speciation of fragranced products is central to understanding how they may impact indoor air in real-world settings.¹ The wider health effects of VOCs indoors are well explored and documented; it is clear that long-term exposure to high concentrations of certain VOCs may carry with it an increased likelihood

of experiencing negative health effects.² In 2019, Public Health England (PHE) released a set of recommended exposure limits for selected VOCs in indoor settings, based on literature data collated by Shrubsole *et al.*^{3,4} Not all VOCs are included in these guidelines, unsurprising given that many thousands of different VOCs exist. For many VOCs there are still some basic uncertainties related to what are representative concentrations indoors, the variability between different indoor spaces, and the emission sources that control concentrations. VOCs such as monoterpenes (including α -pinene and limonene) have particularly complex emissions. They are released from natural sources such as building materials, plants and food products and are also included as fragrance ingredients in many cleaning and personal care products. There is limited information on individual source strengths and a general lack of quantitative assessment of the contributions made to indoor VOCs from individual commercial products. This limits the development of regulatory and/or industry guidance and best practice that might be effective in managing VOC emissions and overall exposure.

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Analytical approaches to sampling VOCs can vary in indoor air studies. Diffusive air sampling using sorbent-packed tubes remains a popular and low-cost option allowing for bulk testing. Sorbent tube sampling followed by thermal desorption (TD) and gas chromatographic (GC) methods remains the international standard for VOC identification in indoor and test chamber facilities, ISO 16000-6: 2021. However, there are limits to the VOCs that can be sampled, which are dependent on the sorbent materials used, as well as potential interferences and reduced species sensitivity from the co-adsorption of water and ozone.^{19,20} Whole air sampling using silica-treated inert canisters is another possible technique and is used in this study. This uses a canister evacuated to vacuum, then exposed to either the test chamber, target room, or connected to a sampling gas line. With sorbent tubes air can enter through natural diffusion or be actively pumped. In vacuum-treated canisters air enters due to a differential in pressure, often through a fixed flow or critical orifice restrictor. Active sampling, either onto tubes, or using vacuum-treated inert canisters generally requires more expensive equipment however it offers a quantitative route to measuring the most volatile VOCs.²¹

The effect of ventilation rate (commonly expressed either as air exchange rates (AER) or air changes per hour (ACH)) on VOC concentrations is well established, with the relationship between the two inversely proportional; as AER increases, indoor VOC concentrations decrease for constant indoor emission rates.^{6,27} The relative influence of variable ventilation rates compared with varying usage of VOC-containing products is uncertain, in large part because individual activity- and product-based emission rates are not well defined and AER is building- and occupant-specific. There has been some investigation of the relationships between temperatures, sink effects and ventilation rates on VOC concentrations, with results showing a dominance of ventilation over temperature and sink effects for short periods.²⁸ The measurement of AER in real settings requires key information for its calculation, such as internal and external carbon dioxide mole fractions/ppm, room occupancy, room volume, as well as CO₂ generation within the room.²⁹ This data can prove difficult to obtain for large scale residential indoor air studies, which necessitates often the use of proxy indicators for AER.

CO₂ has long been used as an indicator for ventilation rates, both in terms of indoor pollutant exposure and risk to airborne-disease infection, which has been particularly prominent during the SARS-CoV-2 pandemic.^{23,30,31} The work of Jia *et al.* (2021)³² demonstrated a link between CO₂ and total VOC concentration in highly-occupied university lecture theatres, as well as a link between room occupancy levels and TVOC concentrations. This is potentially helpful as it means AER, CO₂ concentrations and TVOC may be inferred from one another, if only a subset of these data were available. We note however that the terminology and definition of TVOC is problematic in so much as it is operationally defined by the method of measurement and is rarely calibrated or directly comparable between studies. Whilst there is no single definition of what TVOC

means, it remains a widely quoted metric in indoor air studies, and despite the lack of comparability it has utility as a means for any study to express the totality of VOC behaviours. The methods used in this paper differ from Jia *et al.*, who used photoionisation detectors (PID) to evaluate TVOC concentrations, and nondispersive infrared (NDIR) sensors to measure CO₂ concentrations continuously. Here TVOC concentrations are calculated by summing all individually measured VOCs, and CO₂ determined through using laser absorption spectroscopy, both based on a 3 day sampling period. While the method in this paper does not show real-time changes in atmospheric VOC composition related to changes in room occupancy, activities or product use, it does give a representative insight into summed VOC/CO₂ concentrations over time, and it illuminates whether the relationship shown by Jia *et al.* is seen in residential settings using an alternative methodology. A study by Murakami *et al.* (2019)³³ showed a weak positive correlation between TVOC and CO₂ concentrations in classrooms with air conditioning, however, the TVOC definition in the work of Murakami *et al.* differed again to that of Jia *et al.* These differences in measurement methods and operational definitions of TVOC does make comparison between studies difficult.

1.1 Objectives

The purpose of this study was to identify in real-world home settings the incremental impacts on ambient indoor VOC concentrations arising from using a single well-controlled VOC-emitting consumer product with known emission rate. The study was conducted across a group of 60 homes in Ashford, UK. Having established a baseline set of indoor VOC observations in all homes,⁵ a commercially available plug-in diffuser was used releasing VOCs from a modified commercial fragrance formulation. The study split homes into those that would use the product (switched on), and those where it was switched off, the latter acting as a parallel control group. The use throughout of a control group was to account for any systematic and large-scale changes to ventilation rates arising from regional scale meteorological effects acting across the cohort as a whole, for example high windspeeds systematically increasing AER or high/low temperatures leading to changes in window opening. The study explored the relationships between use of the plug-in diffuser, observed VOCs (both individual VOCs and summed VOC amount), indoor CO₂ mole fraction (as a proxy for AER) and wider self-reported chemical product use statistics from occupants in the homes.

2. Methodology

2.1 Experimental methodology

Homes were selected from a pre-existing non-trained fragrance industry panel cohort based in the Ashford area, UK. Each homeowner was given a commercially available mains-powered fragrance diffuser (referred to as 'the diffuser') with known liquid fragrance formulation. The fragrance was contained in a dipropylene glycol carrier liquid. Homes were split into two groups of 30 homes, each group studied over a four-to-five-week

period. The first 30 homes were sampled between October and December 2021, and the second 30 homes between January and March 2022. Sampling was into 6 Litre stainless-steel canisters internally treated with silica (Restek, PA, USA and Entech, CA, USA) attached to a flow restrictor to limit sample into the canister to 1 Litre every 8 hours ($\sim 2 \text{ mL min}^{-1}$). Heeley-Hill *et al.* found these flow restrictors to sample linearly over a 48 hour period, with a reduced rate between 48 hours and 72 hours, achieving the full 6 L sample at ambient pressure after 72 hours.⁵ Canisters were tagged with unique identifiers evacuated to at least 170 Pa then sent for dispersal among the participants. Samples were taken over weekend periods, starting at 7 am on Friday and finishing at 7 am the following Monday. Filled samples were then returned for processing and analysis. Canisters were then evacuated once again, and the process repeated. Personal details relating to the participants and their homes were fully blinded to the University of York and anonymity of the participants maintained throughout the study.

At least four consecutive weekend samples were taken from each home:

- (1) an initial sample with the diffuser switched off (baseline sample),
- (2) a sample taken after the diffuser had been switched on at the start of the in-use period,
- (3) a further sample taken a week later from (2) where the diffuser remained switched on throughout and,
- (4) a final sample taken several days after the diffuser had been switched off (post-diffuser sample).

This gave as a minimum two 'diffuser off' samples and two 'diffuser on' samples for each home. 15 houses from each group provided a second post-diffuser sample two weeks after the diffuser was switched off.

On return to the lab the canisters were pressurised and thus diluted to 1 bar (gauge pressure) with highly purified air, free of VOCs. This diluent air was purified by passing through a bed of platinum beads heated to 400 °C to fully oxidise any VOCs present. The same purified air was directly analysed by the instrument in order to quantify any impurities or interfering compounds contained therein (none were found). Samples were analysed using a custom thermal desorption unit (TDU) coupled to an Agilent 7890A gas chromatograph (GC) fitted with flame ionisation detectors (FID) and an Agilent 5977A quadrupole mass spectrometer (QMS) manufactured by Agilent Technologies, CA, USA (GCFID-QMS). A flow chart for the system can be seen in ESI Fig. 1.†

Sample canisters and blank samples were connected to the TDU through a 16-port Valco microelectric actuator multi-position valve (VICI Valco Instruments Co. Inc., TX, USA). 500 mL was withdrawn from each sample and first passed through a water trap, comprising a 30 cm length of 1/16" silica-coated stainless-steel tube held at -40 °C , which removed moisture from the samples. The dried samples were then passed through a pre-concentration trap comprising a 30 cm length of 1/16" silica-coated stainless-steel tube packed with both Carbopack™ X 40–60 mesh and Carbopack™ B 60–80 mesh adsorbents (Sigma-Aldrich, subsidiary of Merck KGaA,



Canisters were randomly selected for blank canister tests to ensure there was no carry-over between samples, and blank tests (using the highly purified air used to dilute canister samples) were interspersed in the sample sequence, along with NPL 30 calibrations. A typical sequence would see a group of 15

Occasionally chromatograms obtained from samples would present with disrupted elution of ethane, ethene and propane peaks, shown in ESI Fig. 3(a).[†] This effect was seen multiple times in each batch of 15 canisters, anecdotally associated with high CO₂ mole fractions. When this occurred, samples were re-run at lower canister pressures until a good chromatogram was produced. ESI Fig. 3(b)[†] shows the same sample canister which produced the data in ESI Fig. 3(a),[†] but after being reanalysed. While this effect added time it was not detrimental to the overall data capture. A small number of samples ($n = 3$) were discounted due to canisters not having taken a full sample (indicated by the canister not being at ambient pressure upon arrival at University of York). This was likely due to participants (all from the same home) not successfully opening the sampling valve. A small number of samples ($n = 2$) were deemed spoiled as the canisters had been exposed to a leak from the diffuser during transport. After these samples were removed, this left a total of $n = 259$ individual useable samples. An image of the plug-in diffuser used in this study is shown in ESI Fig. 4.[†]

All data processing and analysis was completed using the R programming language (v.4.1.3), through RStudio Software. Data manipulation was completed using packages contained within the *Tidyverse* collection of packages, mainly *ggplot2* and *dplyr*. Boxplots show values in the order of (from bottom-to-top): lower outliers, 5th percentile, 25th percentile, median value, 75th percentile, 95th percentile, and upper outliers, along with mean values displayed as a red diamond on select boxplots.

Statistical testing was completed using Wilcoxon signed-rank tests, using a confidence interval of $\alpha = 0.95$. Statistical significance using this approach was defined as a p -value yielded from a Wilcoxon signed-rank test lower than 0.05. Unless stated otherwise in the paper, it should be assumed this is our threshold for significance for all statistical testing done. Mean values were not used in data visualisation but were used in statistical testing. Tests were completed using the house the sample originated from as the identifier, so observations were aggregated into groups typically of 1 or 2, which would not yield a difference whether mean or median values were used in statistical testing. Sensitivity tests were completed using median values for statistical analyses however, which confirmed that both methods gave the same conclusions. Important to note is that while a change in species concentration may be of statistical significance, this does not necessarily mean the change in species concentration was large in absolute terms of $\mu\text{g m}^{-3}$. Correlation matrices were produced using the *corrplot* package using visual aids to help show correlations: a forward slanting blue line indicates a positive correlation, a full circle indicates no correlation, and a backward slanting red line indicates a negative correlation. A more intense colour and a narrower line indicates a stronger correlation (both for positive and negative correlations). Covariance analysis was completed through initially rescaling concentration and product use data from 0 to 1. Covariance values were then calculated and assigned, then rescaled from 0 to 100 using *BBMsc*. The matrix was then displayed graphically using the *tile* function in *ggplot2*. Diffuser increment plots were produced using the *raster* and *contour_filled* functions in *ggplot2*, with additional contour lines and contour line labels added using the *contour2* and *geom_text_contour* functions in the *metR* package.

TVOC concentrations were calculated as being the sum of all individually quantified VOC in each sample. These included C_2 to C_8 hydrocarbons, and a range of monoterpenes and oxygenated VOCs (oVOCs) such as esters and alcohols. An additional metric of 'fragrance TVOC' was defined as the sum of individual concentrations for α -pinene, β -pinene, γ -terpinene, benzaldehyde, p -cymene and eucalyptol. Important to note however is that these six species do not reflect the total fragrance formulation, and there were many species within the fragrance which were not detectable in ambient air, owing to limits with the analytical method and often very low compound vapour pressures. For transparency, all data gathered, including survey answers, species concentrations and statistical test results, have been uploaded to the Centre for Environmental Data and Analysis (CEDA) repository.

2.3 Participant panel and survey

Each participant was provided with a tablet-based questionnaire which included questions on property and construction details, occupancy, participant activities and household VOC product use. This is shown in ESI Fig. 5.† Product use information was gathered on a cumulative basis over the 3 day sampling period. Product use statistics were limited to

frequency of use of product types (e.g. paints, air fresheners, sealants etc.), and not brands or sub-types, or absolute quantities. Instructions on use of the sampler (how to open and close the sampling valve, and when) and where to place it were given in written form to the participants. The canisters were placed in the main living space within the house, in most cases the living room. Approximate distances between the diffuser and sampling canisters were reported by each participant. There was a 100% response rate to the survey.

3. Results

3.1 Individual VOCs and TVOC

Table 1 shows the median concentration, 5th percentile, 95th percentile and standard deviation for each of the 47 species, separated into diffuser off and diffuser on values for all 60 homes combined. Running Wilcoxon signed-rank tests were conducted on the aggregated data paired by the house the sample came from. Statistically significant increases (p -values for all following species were less than 0.05) in mean concentration were seen for 20 species, including α -pinene, eucalyptol, ethanol, benzaldehyde and TVOC with the diffuser turned on. There were also statistically significant decreases (p -value for all following species were less than 0.05) in aggregated mean concentrations, again paired by house, when the diffuser was turned on: acetone, δ -terpinene, γ -terpinene and o -xylene. Fig. 1(a) shows TVOC values graphically through boxplots, differentiated by diffuser off and diffuser on values. α -Pinene constituted $\sim 5\%$ of the fragrance formulation and was the most volatile aroma compound by vapour pressure. The diffuser off and diffuser on values for α -pinene can be seen in Fig. 1(b), alongside the emission-estimated diffuser α -pinene increment. This increment is elaborated on later in the paper and compared to estimated values based on emissions rates, AER and room sizes. To confirm that the diffuser was delivering a detectable fragrance, occupants were asked to note whether there was a perceivable odour. 55 homes had at least one occupant report a detectable fragrance attributable to use of the diffuser. For occupants that reported being unable to detect the fragrance, diffuser operation was alternatively confirmed by a negative mass-change (range of 5 g to 7 g of liquid fragrance lost), indicating that the lack of perceived fragrance may have arisen due to those individuals' odour detection response, or other factors such as a high AER, or presence of other more dominant fragrance-emitting sources. All diffusers used in the study were confirmed to work before deployment. Statistical mean weight loss across all diffusers with available data, excluding those which leaked, was 6.84 g with a standard deviation of 1.93 g. Boxplots for other species as well as fragrance TVOC are shown in ESI Fig. 6.†

There are often common patterns for VOCs related to common emission sources, time of day year^{-1} , temperature, building materials and many other variables.^{34–36} To study any correlation between quantified species, correlation matrices were constructed between all quantified species, shown in ESI Fig. 7(a) and (b)† for diffuser off and diffuser on correlations, respectively. Inter-fragrance VOC correlations were generally strengthened when the diffuser was turned on. Most fragrance



Table 1 Median, 5th and 95th percentile and standard deviation values for all quantified VOCs, differentiated by diffuser status. All median and percentile values are expressed in $\mu\text{g m}^{-3}$

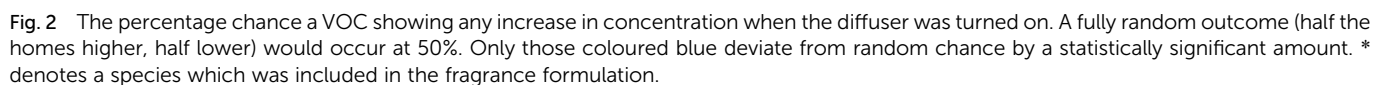
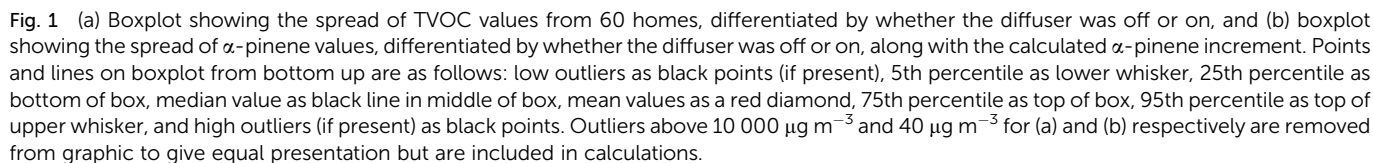
Species	Median concentration		5th Percentile		95th Percentile		Standard deviation	
	Diffuser off	Diffuser on	Diffuser off	Diffuser on	Diffuser off	Diffuser on	Diffuser off	Diffuser on
Ethane	6.9	8.5	1.9	3.4	160	140	100	94
Ethene	1.2	1.3	0.34	0.45	4.1	4	4.8	6.5
Propane	99	150	4.3	5.3	960	1300	330	400
Propene	0.4	0.58	0.13	0.19	1.6	1.7	7.9	11
Iso-butane	120	210	2.1	4.2	1200	1300	380	390
<i>n</i> -Butane	230	310	6.1	9.3	1300	1300	460	500
Acetylene	0.35	0.47	0.17	0.24	1.3	1.7	0.81	1
But-1-ene	0.15	0.22	<0.1	<0.1	0.61	0.68	0.24	0.37
<i>cis</i> -But-2-ene	0.67	1.6	0.2	0.19	5.6	12	2.4	3.9
Isopentane	4.1	7.8	0.79	1.3	42	48	13	20
<i>n</i> -Pentane	1.4	1.5	0.42	0.61	11	8	5.1	2.6
<i>cis</i> -Pent-2-ene	<0.1	0.1	<0.1	<0.1	0.29	0.45	<0.1	0.18
<i>n</i> -Hexane	0.23	0.29	0.12	0.15	1.2	0.92	0.83	0.28
Isoprene	2.1	2.2	0.49	0.57	6.4	5.7	2	1.9
<i>n</i> -Heptane	1.1	0.42	<0.1	0.12	6.4	3	2.3	1.1
<i>n</i> -Octane	0.21	0.29	<0.1	0.11	1.6	4.7	1.3	1.9
Ethylbenzene	0.63	0.61	<0.1	0.12	3.4	4.8	6.7	1.9
<i>m</i> -Xylene	0.62	0.73	0.17	0.21	2.7	4.5	4.5	2.1
<i>o</i> -Xylene	1.5	0.68	0.18	0.1	23	4.4	8.1	19
1,3,5-Trimethylbenzene	0.17	0.25	<0.1	<0.1	0.94	2.9	1	1.7
1,2,4-Trimethylbenzene	0.76	1.1	0.28	0.2	3.1	3.8	1.6	1.9
1,2,3-Trimethylbenzene	0.16	0.25	<0.1	<0.1	1.3	1.1	0.95	0.83
Benzene	0.46	0.44	0.15	0.15	1.2	1.3	0.42	0.43
Toluene	2.4	2.5	0.72	0.8	11	11	3.2	4.6
Styrene	0.18	0.16	<0.1	<0.1	0.82	0.47	0.47	0.16
Acetone	36	36	13	14	150	130	81	150
Acetaldehyde	11	11	3.7	3.8	28	24	52	7.3
Hexanal	5.3	5	<0.1	<0.1	13	14	4.4	4.5
Butan-2-one	4	4.5	0.85	0.86	21	31	8.5	11
Methanol	48	49	16	11	160	170	49	58
Benzaldehyde	0.17	0.18	<0.1	<0.1	0.36	0.38	<0.1	0.11
Ethanol	730	1000	130	250	3100	2700	960	980
Ethyl acetate	5.3	4.6	0.43	0.48	50	56	39	23
Butyl acetate	1.4	1.5	0.21	0.15	20	18	20	15
Propyl acetate	1.9	2.1	0.15	0.21	14	50	10	20
Acetonitrile	7.5	6.1	1.6	0.32	16	37	10	18
Dichloromethane	0.35	0.41	0.11	0.15	2.6	3.4	1.5	2.3
α -Pinene	5.8	8.7	1.2	2.5	32	27	12	8.8
β -Pinene	1.5	1.8	0.4	0.46	8.2	5.3	4.8	1.9
D-Limonene	8.9	10	1.4	1.6	38	40	14	18
Eucalyptol	<0.1	<0.1	<0.1	<0.1	0.39	0.53	0.62	0.18
β -Terpinene	0.12	<0.1	<0.1	<0.1	0.64	0.64	0.46	0.23
γ -Terpinene	0.3	0.26	<0.1	<0.1	2.5	1.8	0.99	0.53
δ -Terpinene	<0.1	<0.1	<0.1	<0.1	0.82	0.58	0.32	0.2
3-Carene	0.63	0.61	0.13	0.13	3.3	2.7	1.2	1.4
β -Myrcene	0.18	0.19	<0.1	<0.1	0.72	0.7	0.25	0.46
<i>p</i> -Cymene	0.78	0.68	0.23	0.24	2.9	2.5	1	0.84

VOCs exhibited a positive correlation with each other both with the diffuser off and when on. There were mostly positive correlations between other non-methane hydrocarbons (NMHCs) C_2 to C_6 , seen most strongly between C_2 to C_4 NMHCs. Species which show stronger correlations with each other mostly arise from similar sources, for example C_3 and C_4 hydrocarbons commonly derive from aerosols, monoterpenes from natural and fragranced products, and monoaromatic

compounds from solvents and combustion of petroleum-derived fuels.^{37–39}

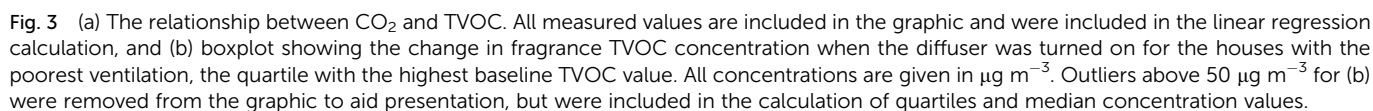
To assess the changes in individual species concentration when the diffuser was on *versus* off, Wilcoxon signed-rank tests were performed. These tests were initially conducted on aggregated values, differentiated by diffuser status and statistics were paired by the house the sample was taken from. Individual species observations are shown in ESI Fig. 8.† Taking





Each participant self-reported the distance between the diffuser and the sampling canister, specific to each sample taken. To

72 hour averaged CO₂ mole fraction was quantified from a subset of houses ($n = 28$) in the second cohort, ranging from 482 ppm to 2019 ppm. CO₂ mole fraction was compared against TVOC values from the same sample, shown in Fig. 3(a). Lines of regression were calculated using a linear model, with R and p values being calculated using Pearson's R. A positive correlation between CO₂ fraction and TVOC concentration is seen in both



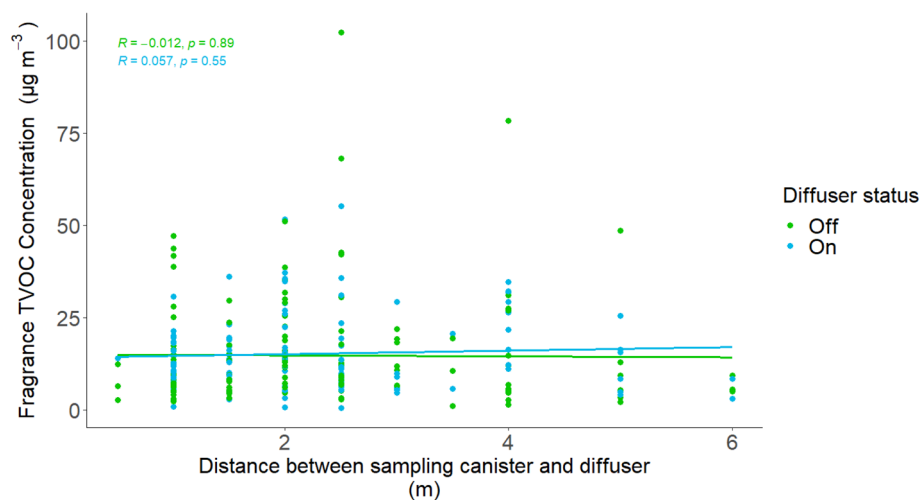


Fig. 4 The relationship between the sum of fragrance VOCs and the distance between the sampling canister and the fragrance diffuser. Regression statistics were calculated using a linear model.

observe how well the fragrance diffused into the airspace in the room, the sum of VOCs was plotted against the distance between the sampling canister and the diffuser, shown in Fig. 4. Distances between sampling canister and diffuser ranged from between 0.5 m to 6 m. No relationship was observed with proximity of sampler to diffuser, implying the diffuser emission was well mixed in the room.

Cumulative product use statistics, limited to product type and number of uses within the 72 hour period, were self-reported by occupants. These values were then plotted against TVOC concentration shown in Fig. 5(a). Total product use was further broken down into the quantiles used previously based on baseline TVOC concentration, and this is shown in ESI Fig. 9.† Fig. 5(b) shows individual product use data for the three-day sampling period, with deodorants being the most used VOC-containing product. To test for any relationship between the frequency of use of particular VOC-containing products and variability observed in individual VOCs, covariance matrices were produced with the four most abundant contributors to TVOC (propane, *n*-butane, iso-butane and ethanol) along with

three fragrance species, α -pinene, β -pinene and eucalyptol. α -Pinene contributed the most to the fragrance formulation; β -pinene is typically used in other fragranced products along with α -pinene, and eucalyptol was included owing to its relatively low concentrations found in the baseline samples. Deodorant, cleaning sprays and aftershave were chosen as the products for comparison since they were the most frequently used. Plug-in diffuser was included as it was known that the use value for this product use increased by exactly 1 per day during the 'diffuser on' sampling period. The matrices produced are seen in Fig. 6(a) for the diffuser off matrix, and Fig. 6(b) for the diffuser on matrix. It is clear that few significant relationships exist between frequency of usage of individual products and variability in specific VOC concentrations indoors.

3.4 Bottom-up estimates of increments in concentration

Diffusers were chosen for this study because their emission rates are well controlled and are not affected by user behaviour. By weighing each diffuser before and after use the mass of fragrance and VOC emitted can be quantified. If a constant rate

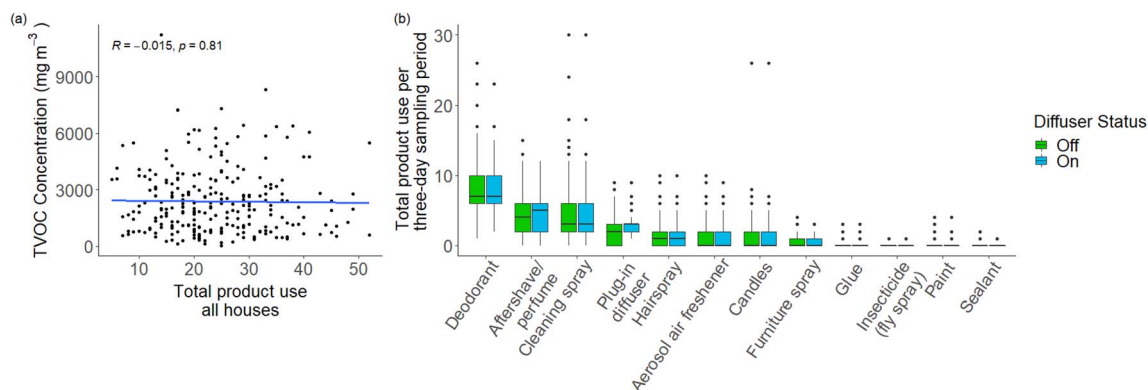


Fig. 5 (a) The sum of all measured VOCs ('TVOC') against total product use frequency from all houses, and (b) the frequency of product use in each home over each three-day sampling period, differentiated by diffuser status.



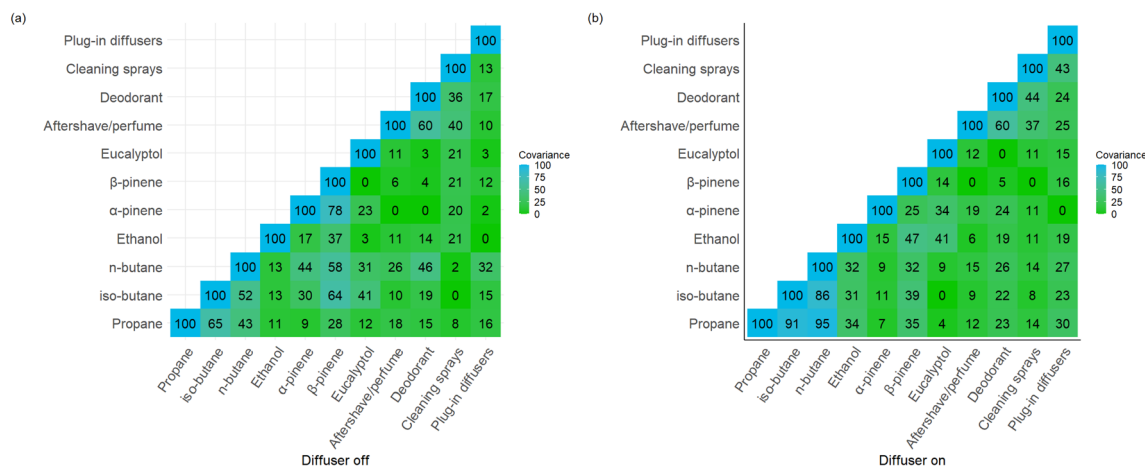


Fig. 6 (a) Covariance matrix between selected VOC species and products for all homes when the diffuser was off, and (b) when the diffuser was turned on. Product use statistics and species concentrations were rescaled on a 0 to 1 scale prior to covariance calculations being completed. Covariance values were then rescaled from 0 to 100.

of loss is assumed, then this can be converted into an emission rate in units of $\mu\text{g h}^{-1}$. An assumption is made that the profile of VOCs emitted from the diffuser does not substantially change over the 72 hour period (or indeed over the study period). Industry norms of technical performance for products of this type would aim to meet this expectation in order to deliver a consistent fragrance to consumers.

Using weight loss data for each diffuser along with the fragrance formulation, the emission rate for each fragrance VOC could be calculated. Additionally, for samples with known CO₂ mole fractions, AER was calculated as an expression of air changes per hour (ACH). These values are shown in Table 2. To calculate this metric, some conservative assumptions were made: where the volume of the room the sample was obtained from was not reported by the homeowner, a room volume of 30 m³ was used derived from data available from the Royal Institution of British Architects (RIBA);⁴⁰ that the room experienced air exchange only with (cleaner) outdoor air and not the rest of the house; external CO₂ mole fractions were assumed to be 450 ppm for all samples; natural CO₂ generation was estimated to be 0.46 L per min per person, as per the work of Batterman (2017),²⁹ and the recorded number of occupants in the house were present in the room being

sampled. ACH was calculated using a one-compartment box model. This was justified as Fig. 4 indicated that there appeared to be no ‘personal cloud’ of higher fragrance concentrations in the immediate proximity to the diffuser. ACH was calculated according to eqn (1):

$$A_H = \frac{6 \times 10^4 n G_P}{V(C_{\text{in}} - C_{\text{ex}})} \quad (1)$$

where A_H is air changes per hour (h^{-1}), n is the number of room occupants, G_p is natural CO_2 generation per person (L min^{-1}), V is room volume (m^3), C_{in} is the internal CO_2 mole fraction within the room being sampled (ppm), and C_{ex} is the external CO_2 mole fraction (ppm). Once ACH was calculated, this then allowed for the calculation of the contribution the diffuser had to elevating a fragrance VOC concentration in a model room. This can be calculated using eqn (2):

$$C = \left(\frac{q}{A_H V (1 - e^{-A_H t})} \right) 10^6 \quad (2)$$

where C is concentration of the species ($\mu\text{g m}^{-3}$), and q is emission rate of the species (g h^{-1}). All other variables remain the same as previously stated. However, as t approaches infinity, the calculation can be simplified, shown in eqn (3):

Table 2 Median, 5th and 95th percentile and standard deviation values diffuser output rates, expressed in $\mu\text{g h}^{-1}$, and for fragrance species increment concentrations, expressed in $\mu\text{g m}^{-3}$

	Diffuser output ($\mu\text{g h}^{-1}$)				Fragrance species concentration increment ($\mu\text{g m}^{-3}$)			
Species	Median	5th Percentile	95th Percentile	Standard deviation	Median	5th Percentile	95th Percentile	Standard deviation
α -Pinene	1270	680	1830	360	4.69	1.1	17.4	7.15
β -Pinene	24.5	13.2	35.4	6.96	0.0907	0.0213	0.336	0.138
Benzaldehyde	39.6	21.2	57.2	11.2	0.265	0.0624	0.982	0.404
<i>p</i> -Cymene	5.42	2.91	7.83	1.54	0.0201	0.00472	0.0743	0.0306
Eucalyptol	71.6	38.4	103	20.3	0.146	0.0345	0.543	0.224
γ -Terpinene	2.83	1.52	4.08	0.803	0.0105	0.00246	0.0388	0.016

$$C = \left(\frac{q}{A_H V} \right) 10^6 \quad (3)$$

Eqn (3) was used in the calculation of fragrance VOC concentration increase within the modelled room. The bottom-up estimated increment in concentrations shown in Table 2 fall broadly within the range of observed α -pinene concentration increments, shown as the third boxplot in Fig. 1(b). It should be recognised that these increment values do not account for simultaneous removal through oxidative reaction pathways or any surface loss of fragrance species into VOC reservoirs or sinks.

3.5 Diffuser increment estimates

Using a variety of room volumes and ventilation rates, as well as a constant diffuser output rate based on values from this study, a spread of diffuser increments in VOC concentrations can be calculated for all combinations of room volume and ACH.

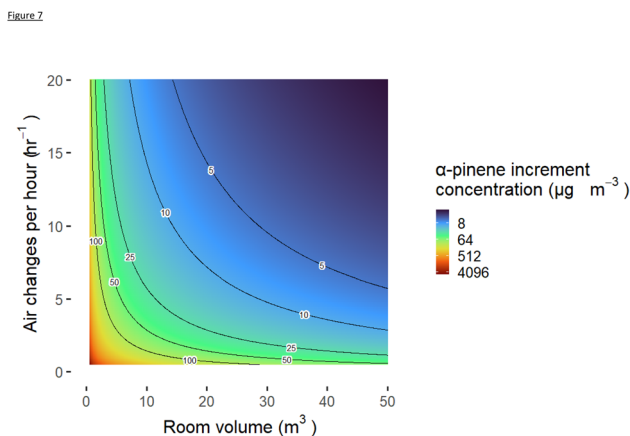


Fig. 7 A raster plot overlaid with contour lines showing the α -pinene increment expected for different room volumes and ventilation rates. A continuous colour scale was applied to α -pinene increment values, and the scale was transformed on a \log_2 scale to aid visualisation.

Plotting these values with room volume on the x-axis, ventilation rate on the y-axis, and with a colour gradient of species concentration as a z-axis variable, a contour plot can be produced for any species included in the diffuser formulation. The contour plot of increments in α -pinene from a single diffuser are shown in Fig. 7. Ventilation rates and room volumes were capped at 20 h^{-1} and 50 m^3 respectively, with the vast majority of physically plausible combinations of AER and room sizes experiencing a predicted increment of no more than $100 \mu\text{g m}^{-3}$. Whilst this model cannot replicate exactly homes tested in the real world, the bottom-up calculation is helpful in demonstrating that the modest increases in VOCs such as α -pinene that were measured when a single diffuser was used are of the same order of magnitude as would be predicted using a bottom-up methodology. For reference the α -pinene 1 day exposure limit was recommended at $4500 \mu\text{g m}^{-3}$ by PHE; this does not however necessarily constitute a long-term exposure recommendation. A single diffuser used in a typical domestic room volume of 30 m^3 (RIBA average room and house sizes for south-east England) and a recommended minimum ventilation rate of 6 ACH would be predicted to give rise to a room increment of around $10 \mu\text{g m}^{-3}$ in α -pinene, again broadly in line with the observations. We note that the loss of VOC from a room is assumed to be solely from ventilation and dilution and that oxidation or deposition are ignored, hence Fig. 7 is likely to represent an upper bound of increment.

3.6 Generation of secondary oxidation products from diffuser use

Carslaw and Shaw (2019) described an operational metric to indicate the potential of a VOC to undergo further reactions within indoor airspaces and create secondary products, named the secondary product creation potential (SPCP).⁴¹ The SPCP is defined as the sum of a range of secondary products created from degradation of a VOC, divided by the mixing ratio of the primary species added, in units of ppb of secondary products per ppb of VOC. It is of relevance for health since secondary



Fig. 8 (a) Calculated total SPCP values differentiated by diffuser status, and (b) SPCP from fragrance species, where data was available. The SPCP * mole fraction product is given as a dimensionless quantity, as SPCP is expressed in units of ppb ppb^{-1} . Outliers below -4 and above $+4$ for (a) and outliers above 1 for (b) were removed from the graphic to aid presentation, but were included in the calculation of quartiles and median values.



and amounts released are not subject to end-user variability. Using a more targeted metric related to secondary product creation potential did not lead to any substantial change in conclusions. Whilst fragrance-related VOCs made up a larger fractional contribution to the SPCP metric there was no statistically significant increase in calculated SPCP across all homes when diffusers were used. A bottom-up model for estimating the incremental increases in a VOC when a single diffuser is used indicated that increments of $>450 \mu\text{g m}^{-3}$ (a high safety margin of 1/10th the 24 hour exposure recommendation) are physically plausible but would require a combination of very small room volumes (below 5 m^3) and very low air exchange rates. It is possible that additional user instruction might provide guidance against use in such unusual situations.

Author contributions

ACL, TW, NO, CJ, BX and GA designed the original experiment. JRH and SJA developed the analytical instruments and methods and TW, JRH and SJA performed the laboratory analysis. TW undertook the data analysis and visualisations and SKG and ACL guided the statistical and data methods. All authors contributed to the writing of the manuscript and the development of its conclusions.

The use of room fragrance products such as plug-in diffusers undoubtedly impacts on the airspace composition within the area the product is used. Clearly using any VOC-containing products in poorly ventilated spaces will result in a higher VOC concentration and higher exposure to the user than would occur in a well-ventilated space. In this study virtually all participants reported that they could detect the presence of fragrance when the diffusers were switched on indicating that in a purely functional sense the devices were working as intended by the manufacturer. However quantitatively determining by how much VOCs were incrementally raised by using the diffuser was not straightforward since the diffuser emitted VOCs into an already highly congested indoor atmosphere containing many VOCs at higher concentrations, including some species that were also in fragrance formulation. Future inclusion of appliance use data in surveys may prove insightful, as the frequency of use of appliances such as gas fires may help with interpreting the presence of high CO₂ fractions that may not be from respiration.

When the study homes were segmented by air exchange rate, inferred from concurrent CO₂ measurements acting as a proxy, then a small increment in VOCs could be detected. For homes in the lowest ventilation quartile, α -pinene concentrations increased on average from 9 to 14.5 $\mu\text{g m}^{-3}$, with the median value increasing from 6 $\mu\text{g m}^{-3}$ to 14.1 $\mu\text{g m}^{-3}$. This increment in α -pinene was set against a background mean TVOC concentration in that quartile of >3500 $\mu\text{g m}^{-3}$. By using the weight loss of fragrance from the diffuser, a bottom up estimate of increment increase in α -pinene was calculated to be around 10 $\mu\text{g m}^{-3}$, assuming 6 ACH, a typical room volume of 30 m³ and no oxidative/depositional losses. In this regard the use of a fragrance diffuser for studies in real-world settings is helpful since the emission rate is well controlled, formulation is known,

NO, CJ, and GA are employees of Givaudan UK Ltd, Givaudan Fragrances Corp. and BX is an employee of Bath & Body Works, Inc. who are industrial suppliers of chemicals and finished household and personal care products. To support independence all analytical work and data analysis was undertaken by University of York and no restrictions placed on freedoms to publish. To support transparency in research, data collected in this study is freely available from the Centre for Environment Data and Analysis (<https://www.ceda.ac.uk>).

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