

Combined application of Online FIGAERO-CIMS and Offline LC-Orbitrap MS to Characterize the Chemical Composition of SOA in Smog Chamber Studies

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Abstract. A combination of online and offline mass spectrometric techniques was used to characterize the chemical composition of secondary organic aerosol (SOA) generated from the photooxidation of α -pinene in an atmospheric simulation chamber. The filter inlet for gases and aerosols (FIGAERO) coupled with a high-resolution time-of-flight iodide chemical ionization mass spectrometer (I-ToF-CIMS) was employed to track the evolution of gaseous and particulate components. Extracts of aerosol particles sampled onto a filter at the end of each experiment were analyzed using ultra-performance liquid chromatography ultra-high-resolution tandem mass spectrometry (LC-Orbitrap MS). Each technique was used to investigate the major SOA elemental group contributions in each system. The online CIMS particle-phase measurements show that organic species containing exclusively carbon, hydrogen and oxygen (CHO group) dominate the contribution to the ion signals from the SOA products, broadly consistent with the LC-Orbitrap MS negative mode analysis which was better able to identify the sulphur-containing fraction. An increased abundance of high carbon number ($nC \geq 16$) compounds additionally containing nitrogen (CHON group) was detected in the LC-Orbitrap MS positive ionisation mode, indicating a fraction missed by the negative mode and CIMS measurements. Time series of gas-phase and particle-phase

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oxidation products provided by online measurements allowed investigation of the gas-phase chemistry of those products by hierarchical clustering analysis to assess the phase partitioning of individual molecular compositions. The particle-phase clustering was used to inform the selection of components for targeted structural analysis of the offline samples. Saturation concentrations derived from near-simultaneous gaseous and particulate measurements of the same ions by FIGAERO-CIMS were compared with those estimated from the molecular structure based on the LC-Orbitrap MS measurements to interpret the component partitioning behaviour. This paper explores the insight brought to the interpretation of SOA chemical composition by the combined application of online FIGAERO-CIMS and offline LC-Orbitrap MS analytical techniques.

1. Introduction

Secondary organic aerosol (SOA) makes a significant contribution to atmospheric aerosols, which have an important influence on climate and adverse impact on human health and air quality (Hallquist et al., 2009; Kroll and Seinfeld, 2008; Nel, 2005). It is important to understand the formation, composition and behaviour of SOA due to their contribution to many important atmospheric processes, such as cloud formation (Hallquist et al., 2009). There are more than 10,000 individual organic species in the atmosphere and those almost entirely in the vapour phase. Volatile organic compounds (VOCs) are known to be important in tropospheric ozone and SOA formation (Goldstein and Galbally, 2007). Biogenic volatile organic compounds (BVOCs) emitted from plants (such as monoterpenes) and anthropogenic volatile organic compounds (AVOCs) (such as aromatics), undergo atmospheric oxidation processes to generate SOA, leading to the formation of a large number of oxidized products (Goldstein and Galbally, 2007; Guenther et al., 2012; Hoyle et al., 2009; Iinuma et al., 2010; Mcfiggans et al., 2019). As a result of the various VOCs involved and the complexity of the oxidation processes, oxidation products span a wide range of molecular composition and physiochemical properties (Mentel et al., 2015; Ma et al., 2008; Mohr et al., 2019; Mutzel et al., 2015; Laj et al., 2009). Such chemical complexity poses a major challenge in the molecular characterization of SOA.

Atmospheric simulation chambers play a key role in the study of chemical processes that affect the composition of the atmosphere, enhancing mechanistic understanding of atmospheric chemistry at a molecular level (Cocker et al., 2001; Rohrer et al., 2005; Wang et al., 2011; Burkholder et al., 2017; Ren et al., 2017). In recent years, new developments in the

70 techniques and design of laboratory experiments have been carried out to investigate the
chemical composition of SOA generated from the biogenic and anthropogenic sources under
various well-controlled conditions, exploring the particle properties, identifying tracers and
SOA formation pathways (Winterhalter et al., 2003; Ng et al., 2006; Nguyen et al., 2010;
Chhabra et al., 2011; Zhang et al., 2015; Schwantes et al., 2017; Lopez-Hilfiker et al., 2014).
Although some progress has been made in characterising sources and formation pathways
75 from some biogenic or anthropogenic precursors, there are still significant gaps remaining in
our understanding of the chemical properties of SOA.

Since there is currently no perfect instrument capable of providing detailed chemical
characterisation of SOA (Hallquist et al., 2009; Calvo et al., 2013), various analytical
techniques have been established to better characterize SOA chemical composition. Mass
80 spectrometric techniques have been widely employed both online and offline, using a wide
range of sample introduction techniques, ionization methods, and mass analyzers (Hallquist
et al., 2009; Nizkorodov et al., 2011; Laskin et al., 2012). Liquid chromatography coupled
with electrospray ionization orbitrap mass spectrometry (LC-Orbitrap MS) can probe the
chemical composition of individual polar and non-polar particle-phase products with accurate
85 mass measurement (Perry et al., 2008; Banerjee and Mazumdar, 2012; Mutzel et al., 2021).
LC-MS has been extensively used for the chemical characterization of organic aerosols,
quantifying targeted tracers and characterising using tandem MS to determine the molecular
identity of ambient SOA (Samy and Hays, 2013; Inuma et al., 2007; Parshintsev et al., 2015;
Chen et al., 2020; Hamilton et al., 2013) and laboratory-produced SOA (Winterhalter et al.,
90 2003; Pereira et al., 2014; Mutzel et al., 2021; Mehra et al., 2020b).

Online collection of aerosol samples followed by fast and automated analysis has more
recently been developed to overcome some limitations of offline measurements (such as
sampling time integration, evaporation, adsorption and potential filter contamination during
transportation or storage). One such technique, the Filter Inlet for Gases and AEROSOLS
95 combined with a soft chemical ionization method and mass spectrometer (FIGAERO-CIMS;
Tofwerk A.G. / Aerodyne Research Inc.), collects filter samples for subsequent thermal
desorption to investigate the SOA particle-phase chemical composition (Lopez-Hilfiker et al.,
2014). The FIGAERO-CIMS has been coupled to different reagent ions to study the chemical
properties of aerosols such as the evolution of gas or particle-phase composition and the
100 volatility of components in laboratory studies and fieldwork (Wang and Hildebrandt Ruiz,

2018; Stark et al., 2017; Bannan et al., 2019; Lee et al., 2020; Lopez-Hilfiker et al., 2016b; Hammes et al., 2019; Thornton et al., 2020). Other online instruments such as particle into liquid sampler (PILS) coupled with ion chromatography (Bateman et al., 2010) and chemical analysis of aerosol online (CHARON) particle inlet coupled to proton-transfer-reaction time-
105 of-flight mass spectrometry (PTR-TOFMS) instruments (Gkatzelis et al., 2018) can also be employed for collection and near real-time analysis of chemical compounds in aerosols. Recently, multiple online analytical techniques have been combined to investigate the chemical properties of atmospheric organic carbon, such as the combination of CIMS, PTR-MS and aerosol mass spectrometer (AMS) (Isaacman-Vanwertz et al., 2017; Isaacman-
110 Vanwertz et al., 2018). There are notable overlaps and differences in the chemical information provided from those online measurements. Nevertheless, few studies have combined offline LC-Orbitrap MS and online FIGAERO-CIMS to study the molecular composition of SOA (Mehra et al., 2020a). By combining both techniques, the detailed molecular composition can be obtained with time evolution/profiling, offering enhanced
115 mechanistic insights.

The purpose of this paper is to explore the benefits of combinatorial LC-Orbitrap MS and FIGAERO-CIMS analytical techniques to investigate SOA chemical composition, demonstrating the power of this combination from a technical perspective. To show this experimentally, α -pinene ($C_{10}H_{16}$), as one of the most abundant monoterpenes (emitted in
120 substantial amounts by vegetation, e.g., many coniferous trees, notable pine) was selected to generate SOA in the presence of NO_x in the Manchester Aerosol Chamber (MAC). First, both online and offline analyses were used to retrieve the major elemental group contributions to the SOA in the α -pinene system. Second, temporal profiles of individual components in gas and particle phases provided by online measurement enabled observation
125 of the evolution of identified components. Third, hierarchical clustering analysis (HCA) was employed to reduce the dimensionality of the complex composition measurements by grouping compounds with similar chemical properties. Finally, saturation concentrations of targeted compounds derived from the partitioning theory in the FIGAERO-CIMS were compared with those derived from the structural information determined using the LC-
130 Orbitrap MS measurements.

2. Instruments and methodology

2.1 Chamber description and experimental design

Experiments were performed in the Manchester Aerosol Chamber located at the University of Manchester to investigate SOA formation using a broad suite of instrumentation. A full description of the MAC and its characterization can be found in Shao et al. (2022). Briefly, the chamber has a volume of 18 m³ (3m (L) × 2m (W) × 3m (H)) and consists of a fluorinated ethylene Teflon bag which is suspended in the enclosure and supported by three rectangular aluminum frames. The central frame is fixed, while the upper and lower ones are allowed to move freely to expand and collapse the chamber when sample airflow is introduced to or extracted from the chamber. Relative humidity (RH) and temperature were controlled by the humidifier and air conditioning and were measured at a few points throughout the chamber by a dewpoint hygrometer and a cross-calibrated capacitive sensor. Two 6 kW Xenon arc lamps and 112 halogen lamps (Solux 50W/4700K, Solux MR16, USA, 16 ×7 rows) are mounted on the wall of the enclosure to provide uniform illumination. The experiments reported here used two arc lamps with quartz glass filters in front of them and 5 rows of halogen lamps to simulate the solar spectra over the wavelength range of 290-800 nm (Alfarra et al., 2012). The reported calculated photolysis rate of NO₂ (j_{NO_2}), investigated from the steady-state actinometry experiments, was ~ 0.11-0.18 min⁻¹ (Shao et al., 2021).

The purified dry air was supplied to the chamber by the laboratory air through a dryer (ML180, Munters) and three filters (Purafil (Purafil Inc., USA), charcoal and HEPA (Donaldson Filtration) filters). The chamber was cleaned by a series of automatic fill/flush cycles with purified air (3 m³ min⁻¹) before each experiment and by overnight oxidation at a high concentration of O₃ (~1ppm) at end of each experiment. An extra 4-5-hour harsh cleaning experiment under strong ultraviolet photolysis was performed weekly with a high concentration of O₃ (~1ppm) to remove the reactive compounds in the chamber. Liquid VOC precursors (α -pinene in this study; Sigma Aldrich, GC grade \geq 99.99% purity) are fed into the chamber via injecting into a heated glass bulb (max: 80°C), vaporized immediately and flushed into the chamber by continuous high purity nitrogen (ECD grade, 99.998 % purity N₂). The required concentration of NO_x (self-made NO_x cylinder, 10% (v/v) NO₂ and 90% (v/v) N₂) is controlled by the injection of NO₂ from a cylinder into the charge line. Seed particles were generated from aqueous solutions using ammonium sulfate (AS, Puratonic,

99.999% purity) by nebulizing via an aerosol generator (Topaz model ATM 230) and introduced into the chamber. During injections, the seed particles and gases are fed into the chamber with continuous purified air at flowrate $3 \text{ m}^3 \text{ min}^{-1}$, ensuring the rapid and well-mixed in the chamber. The relative humidity in the chamber will be adjusted by the custom-built humidifier which comprises a 50L tank fed with ultra-pure water (resistivity $\geq 18.2 \text{ M}\Omega\text{-cm}$), generating water vapour using an immersion heater that heats the water to $\sim 80^\circ\text{C}$.

In this work, we will focus on the application of offline and online techniques for the chemical characterization of gaseous and particulate components formed from the photo-oxidation of α -pinene in three repeat experiments. To obtain sufficient particulate mass on the filter at end of each experiment, an initial reference concentration of α -pinene was targeted at 309 ppb for each experiment, with an initial VOC: NO_x ratio ranging from 6 to 8. The α -pinene studies were part of a series of experiments investigating SOA formation in mixtures and the rationale behind the experimental design is detailed elsewhere (Voliotis et al., 2021). The O_3 concentration was zero at the beginning of each experiment and no additional O_3 was added during experiments. Ammonium sulfate seed aerosol (1g AS dissolving in 100ml deionized water, $\sim 53 \pm 12 \mu\text{g m}^{-3}$) was added to enable more controllable and uniform condensation of the gas-phase products and suppress their nucleation. VOCs, NO_x and seed particles were injected into the clean chamber. All experiments were performed under similar conditions (temperature $\sim 25^\circ\text{C}$, relative humidity $\sim 50\%$, 6-h photooxidation reactions). The initial OH source was generated through the photolysis of the injected NO_2 . During the experiment, photolysis of ozone and unsaturated VOC ozonolysis (or potentially via OH recycling) can also generate OH source. The irradiance for photolysis of both NO_2 and O_3 is provided by two 6 kW Xenon arc lamps and 5 rows of halogen lamps as described in Shao et al. (2022). A 1-hour stabilization period was utilized for measurement of the background conditions in the chamber after injection of the precursors and seed particles but before illumination of the chamber and the onset of photooxidation. At the end of each daily experiment, aerosol sample filters were collected through the 47mm embedded filter holder located at the flushing line of the chamber by pre-processed filters (Quartz filter: $2.2 \mu\text{m}$ pore size) at the flowrate of $3 \text{ m}^3 \text{ min}^{-1}$, before storage in a freezer at -20°C for subsequent extraction and analysis. Filters were pre-baked at 550°C for 5.5 h to remove potential organic contaminants and stored in the freezer (-5°C) until sampling and analysis. The chemical composition of filters was characterized and expressed as peak area through the offline LC-

[MS analysis. The ratio of the largest residual contaminant that was removed from the sample by the blank subtraction is 188, indicating the much lower loading of the blank filters after the pre-treatment.](#) –Table 1 summarises the initial experimental conditions and instrument availability.

200 **Table 1.** The initial experimental conditions and instrument availability.

Exp No	VOC (type)	Seeds	VOC (ppb)	$\frac{VOC}{NOx}$	Seed concentration ($\mu\text{g m}^{-3}$)	FIGAERO-CIMS availability	LC-Orbitrap MS availability
1	α -pinene	AS	309	8.4	<i>n.a</i>	×	✓
2	α -pinene	AS	309	7.7	60.7	✓	✓
3	α -pinene	AS	309	7.2	88.4	✓	✓

Note: *n.a* represents no data due to the instrumental issues. The reported initial VOC concentrations in this table have a $\pm 15\%$ measurement uncertainty.

205 The repeat experiments were used to evaluate the reproducibility of our experimental system, our data processing and provide a quantification of our confidence in the processes leading to the differences in composition. Data differences can arise from a range of sources: control of experimental conditions, instrumental errors and operational differences, data processing and error propagation. The treatment of the data from the three experiments from the different measurement techniques is described in their respective data processing sections.

2.2 On-line FIGAERO-CIMS

210 2.2.1 Instrumentation and operation

215 We employed a FIGAERO-CIMS (Aerodyne Research Inc.) with iodide (I⁻) as the reagent ion to measure the gas and particle phase oxidized species produced from photooxidation reactions of the experiments. This instrument has been described in detail previously (Reyes-Villegas et al., 2018; Bannan et al., 2019) and was operated at a ~ 4000 (Th/Th) resolving power. [Gas-phase species are sampled via a 0.5 m 1/4" I.D. PFA tubing at 1 standard litre per minute \(SLM\) from the chamber. The gas sampling was performed for 30 mins.](#) ~~Gas-phase species are sampled via a 0.5 m 1/4" I.D. PFA tubing at 1 standard litre per minute (SLM) from the chamber and characterized for 30 mins.~~ Meanwhile, particles are collected simultaneously in a separate FIGAERO inlet on a polytetrafluoroethylene (PFTE) filter (Zefluor, 2.0 μm pore size) at the flow rate 1SLM. After 30-min collection, the filter undergoes thermal desorption (15-min temperature ramp to 200 °C, 10 mins holding time and 8 mins cooling down to room temperature). **Six gas phase and particle phase cycles were carried out during each experiment. The sample blanks were collected by two additional gas**

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and particle-phase cycles before the initiation of photooxidation. The first cycle was
225 conducted in the clean chamber condition before the injection of reactants. This was followed
by the second cycle after all species (VOC, seeds and NO_x) had been injected. The gas-phase
data collected in the first cycle was employed for the gas-phase “chamber background”. The
instrument was flushed with ultra-high purity nitrogen (UHP, 99.999% purity, N₂) for 0.2
min every 2 min during each gas-phase measurement, which acted as the gas-phase
230 “instrument background”. To remove the influence of the seed particles, the particle-phase
data collected in the second cycle were used as the particle-phase background correction. To
remove the influence of the seed particles, the particle-phase data collected in the second
cycle were used as the particle-phase background correction.

2.2.2 Data processing

235 Data analysis was performed on the Tofware package (version 3.1.2) using the Igor Pro 7.0.8
(WaveMetrics, OR, USA) environment (Stark et al., 2015). Mass spectra were mass
calibrated, after which the high-resolution peaks were fit via the multi-peak fitting algorithm.
The exact mass of the multiple peaks is then matched with the most likely elemental formula,
with mass errors less than ± 6 ppm. Assigned peaks were fitted through the iterative peak
240 assignment method outlined in Stark et. al (Stark et al., 2015). In this study, the m/z of the ion
adducts with I- range from 200 to 550 was selected to investigate the oxidation products as
signals in this range contribute more than 80% of the total signal (exclude the reagent ions, I,
IH₂O⁻, I₂⁻ and I₃⁻). All identified molecular formulae were thereafter expressed as C_xH_yO_zN_m.

Substantial challenges remain in quantifying all identified compounds owing to a lack of
245 standards. The total signal of identified peaks in the range of m/z 200-550 (excluded the
reagent ions, I₂⁻: m/z 253.809492 and I₃⁻: m/z 380.713964) was used to normalise the ion
signal, expressing the relative contribution of each identified compound with an implicit
assumption of uniform sensitivity (Isaacman-Vanwertz et al., 2018). The gas- and particle-
phase dataset undergoes post-processing in MATLAB (R2017a) code in different ways. For
250 the gas phase, the averaged gas-phase measurement signal of ion i was corrected by
subtracting the average clean chamber background value and instrumental background value.

For the particle phase, the signal in the first 60 - 90s with relatively low and stable signals
was considered as the instrumental background, enabling interference between the gas and
particle mode switching to be removed (Voliotis et al., 2021). All the measured signals were

255 corrected by subtracting the average background value of each ion, after which the
temperature of ion counts i was integrated to estimate the normalized signal. The thermogram
desorption of one compound may have more than one maximum desorption temperature or
enhanced tailing even during the temperature ramp period. This can be caused by the
fragments from higher molecular weight compounds or the presence of isomers with different
260 saturation concentrations (Lopez-Hilfiker et al., 2015; Lopez-Hilfiker et al., 2016a; Lutz et al.,
2019; Stark et al., 2017). These complex thermogram desorption features for one compound
(e.g. two thermogram profiles or enhanced tailing) can be addressed by a custom nonlinear
least-squares peak-fitting routine (Stark et al., 2017; Lopez-Hilfiker et al., 2016b; Lopez-
Hilfiker et al., 2015). A non-linear iterative curve fitting method was deployed to fit the peaks
265 (O'haver, 2021). First, the highest signal was found corresponding to the first maximum
desorption temperature (T_{max}) of the ion. The peak width was set based on the full width at
half maximum (FWHM) and the shape followed by Gaussian distribution. The number of
desorption peaks, the location and the amplitude of each peak were optimized to fit the total
shape of the desorption profile and used to retrieve the total ions for each peak. Figure 1
270 shows an example of desorption peaks of two compounds, which are consistently measured
as major components of the detected α -pinene derived SOA particles. Typically, the results
show that the second peak has a T_{max} between 20 °C and 90 °C higher than the first peak
which is caused by the thermally decomposed accretion products from higher molecular
weight rather than isomers (Lopez-Hilfiker et al., 2015). Thus, the first desorption peak was
275 considered as the monomer generated from the precursor oxidation, and its related integrated
ion counts were employed for normalized signal and partitioning calculations.

Since a common peak list is used for two experiments, the normalised fractional CIMS signal
for the repeat experiments can be reported as averages along with the associated variability
(as in Figures 2 and 4) or as values from one representative experiment without variability
280 (e.g., Figures 3, 5 and 6) associated with supplementary figures representing another
experiment as appropriate.

2.2.3 Hierarchical clustering analysis

Hierarchical clustering analysis (HCA) is an analytical method that is used to group or cluster
datasets with a large number of individual observations based on the similarity of their
285 behaviour over time. It is used here to investigate the time series of gaseous and particulate
oxidation products measured by CIMS and FIGAERO-CIMS. Generally, there are two main

hierarchical clustering analysis methods: divisive hierarchical clustering (working in a top-down method) which describes if a cluster needs to be split and agglomerative hierarchical clustering (working from the bottom-up) which describes if clusters should be combined
290 (Nielsen, 2016; Bar-Joseph et al., 2001; Müllner, 2011). Compared to divisive hierarchical clustering, agglomerative hierarchical clustering is commonly used to cluster measurements with distinct time-series behaviours from mass spectrometry datasets and describe the degree of similarity between any two measurements, reducing the dimensionality of mass spectrometry datasets and improving understanding of bulk properties of the chemical
295 processes (Sánchez-López et al., 2014; Rosati et al., 2019; Koss et al., 2020; Priestley et al., 2021). HCA is independent of calibration or instrumental sensitivity since it relies on the relative differences between time series shapes. The final number of clusters is decided based on the distance between the objects and is decided by the user. In this study, we illustrate how HCA can be implemented to reduce the complexity of a dataset while retaining the chemical
300 information by investigating the oxidation processes or product properties from the photooxidation of α -pinene.

First, to observe the time-series trend changes of each ion and remove the effect of the differences caused by the absolute signal intensity, all measurements are normalized to the highest signal in each time series of each ion. Second, the Euclidean distance between each
305 pair of points of ion A and ion B was calculated (Eq. 1). The sum of the distance for each pair of observations was considered as the distance metric since this approach is most reproducible and least sensitive to outlier points in the time series (Koss et al., 2020).

$$d_{A,B} = \sum_t \text{abs}(A_t - B_t) \quad \text{Eq 1}$$

Third, the average linkage criterion was selected to determine the distance between sets of
310 measurements as this criterion gave more similar or understandable results. The algorithm starts with the distances between all observations. As a first step, the lowest distance between observations A and B is identified and is assigned a new cluster x . The two observations of A and B are removed from the distance dataset and the new cluster x is added into the dataset. Then the distance set between the new cluster x and the remaining observations (named y)
315 was calculated as the average of the distances between each of n individual members of x and m individual members of y over all points i in cluster x and points j in cluster y , as shown in Eq. 2. The smallest distance between those observations is then found and made up into a new cluster iteratively until only one cluster remains. Finally, the dendrogram tree is

320 deployed to display the relationships between each measurement and cluster. The whole process is carried out in MATLAB (MathWorks, Inc.).

$$d_{x,y} = \sum_{i,j} \frac{d(x_i, y_j)}{m \times n} \quad \text{Eq. 2}$$

2.3 Off-line LC-Orbitrap MS

2.3.1 Sample preparation

325 The sample preparation method described in Bryant et al., (2020) was modified for our experimental conditions. Briefly, 4 ml methanol was added to a cleaned and dried extraction vial containing half a filter cut into small pieces and left at room temperature for 2h before sonicating for 30 mins. The extractant was then filtered through a 0.22 μm filter (Thermo Fisher Scientific) using a syringe (1mL, BD PLASTIC PAK, STERILE) into another sample vial, which was evaporated to dryness using a vacuum solvent evaporator (Biotage, Sweden).
330 The sample was reconstituted in 1ml of 10: 90 methanol: water (optima LC-MS grade, Thermo Fisher Scientific) for analysis.

2.3.2 LC- Orbitrap MS analysis

The samples were analyzed using ultra-performance liquid chromatography ultra-high-resolution mass spectrometry (Dionex 3000, Orbitrap QExactive, Thermo Fisher Scientific).
335 Compound separation was achieved using a reverse-phase C_{18} column (aQ Accucore, Thermo Fisher Scientific) with the following dimensions: 100 mm (length) \times 2.1 mm (width) and 2.6 μm particle size. The column was held at 40°C during analysis and the samples were stored in the autosampler at 4°C. Gradient elution was used, starting at 90 % (A) with a 1-minute post-injection hold, decreasing to 10 % (A) at 26 minutes, returning to the starting mobile
340 phase conditions at 28 minutes, with a 2-minute hold to re-equilibrate the column, and the total running time was 30 minutes for each sample. The flow rate was set to 0.3 ml/min with a sample injection volume of 2 μl . Heated electrospray ionization was used, with the following parameters: capillary and auxiliary gas temperature of 320 °C, sheath gas flow rate of 70 (arb.) and auxiliary gas flow rate of 3 (arb.). The mass spectrometer was operated in
345 negative and positive ionization mode with a scan range of m/z 85 to 750 and mass resolving power of 70000 (Th/Th). Tandem mass spectrometry was performed using higher-energy collision dissociation with a normalized collision energy of 65, 115.

2.3.3 Data processing

Data were analyzed by two complementary approaches. The conventional approach for LC-MS analysis entailed extracting fragments and structural information of targeted compounds using the XCalibur software 4.3. A more recent automated non-targeted approach extracts all detected chromatographic peaks from the sample data set, removing any compounds which do not satisfy a set criterion (see below). The molecular formula is automatically assigned for each compound via a bespoke method designed in the analysis platform, Compound Discoverer version 2.1 (Thermo Fisher Scientific). Full details of the methodology can be found (Pereira et al., 2021). Briefly, molecular formulae assignments were allowed unlimited C, H, O atoms, up to 2 S atoms and 5 N atoms, plus > 2 Na atoms and 1 K atom in positive ionization mode. Compounds with a mass error < 3 ppm and signal-to-noise ratio > 3 , a hydrogen-to-carbon ratio of 0.5 to 3 and an oxygen-to-carbon ratio of 0.05 to 2 were included in the data set. Any compounds detected in the procedural (control sample, *i.e.* blank pre-conditioned filter subjected to the same extraction procedure) and solvent blanks (instrumental blanks) with the same molecular formula and a retention time difference within 0.1 min were removed from the sample data. Furthermore, any compounds detected in the chamber background filter with the same molecular formula and a retention time difference within 0.1 min and the ratio of signal intensities between the sample data and chamber background lower than 3 were removed from the sample data. As with online FIGAERO-CIMS, since quantification for all identified compounds using standards is not possible, all compounds were normalized to the total signal to express the relative contributions, implicitly ignoring differential sensitivity in comparison between the two instruments.

The peak list for the LC-Orbitrap MS is not identical for all experiments since some peaks were not detected in all experiments. The uncertainty as to whether such peaks are truly representative of particle-phase oxidation products is captured by classifying the peaks as “common” (common to all repeat experiments) and “inconclusive” (in one specific experiment or common to two of all experiments) and then by expressing their contribution on average across the experiments with the uncertainty across each in Figure 2. In some cases, the results from one representative experiment are used, such as in Figure 3.

2.4 Elemental grouping of identified compounds

The identified species from FIGAERO-CIMS mass spectra and from the automated method for LC-Orbitrap MS analysis described above were grouped according to their elements. Compounds containing only carbon, hydrogen, oxygen and nitrogen were categorized as the CHON subgroup. Components containing only carbon, hydrogen and oxygen were classed as CHO subgroup. Similarly, compounds containing carbon, hydrogen, oxygen and sulfur or containing carbon, hydrogen, oxygen, nitrogen and sulfur fell into CHOS or CHONS subgroups respectively. Compounds were additionally classed according to the numbers of carbon atoms in their molecular formulae: C₂ to C₇, C₈ to C₁₀, C₁₁ to C₁₅, \geq C₁₆. Given the selectivity and sensitivity of the ionization methods, not all ions were equally observed in the two instruments, and thus only the dominant contribution of identified compounds included in the CHO and CHON groups are discussed in this paper. The minor contribution of CHOS and CHONS groups are not attributed in the FIGAERO-CIMS measurement as it is difficult to identify the two species reliably within the trusted error (± 6 ppm). Previous studies also showed difficulties in the identification of isoprene-derived organosulfate compounds from CIMS measurements unambiguously due to the low mass resolution or thermal desorption of organosulfate compounds (D'ambro et al., 2019; Xu et al., 2016). Compounds in the two groups may be included in the unassigned category or may misattribute their signals to the CHO and CHON for the online measurement. The unassigned fractions in the FIGAERO-CIMS measurements (less than 20% of total signal) represent the peaks that are difficult to identify, caused by either the poor signal-to-noise ratios ($S/N \leq 2$, ~40-50%) or inaccessible formulae within the trusted error.

2.5 Calculation of average carbon oxidation state (\overline{OS}_C)

The average carbon oxidation state is commonly deployed to describe the degree of oxidation within a complex oxidation reaction (Kroll et al., 2011). The elemental ratios of oxygen-to-carbon (O:C), hydrogen-to-carbon (H: C) and nitrogen-to-carbon (N: C) are used to calculate the \overline{OS}_C . For the CHON compounds, we assume nitrogen is in the form of nitrate where the oxidation state of N (OS_N) is +5 if the oxygen number of the molecule is equal to 3 or more than 3. The nitrogen is in the form of nitrite where the OS_N is +3 if the oxygen number of the molecule is less than 3. In the FIGAERO-CIMS, the \overline{OS}_C for the CHOS and CHONS compounds is not considered because of the associated challenges in quantifying S-

containing compounds (see above). In the LC-Orbitrap MS analysis, the total signal fractions of CHONS and CHOS groups make low contributions to the total signal, accounting for $\sim 10 \pm 2\%$ in the common peaklist in both modes and for $12.4\% \pm 2.7\%$ and $28.5\% \pm 2.5\%$ in the inconclusive peaklist for negative mode and positive mode, respectively. It is also impossible to assess if the structure of a compound in the CHONS should be assigned as nitrooxy-OSs with both $-\text{OSO}_3\text{H}$ and $-\text{ONO}_2$ groups or as other combinations (e.g. $-\text{NO}_2$ and $-\text{SO}_3\text{H}$, $-\text{NO}_2$ and $-\text{OSO}_3\text{H}$ etc). Given the two considerations, the calculation of $\overline{\text{OS}}_C$ for the CHOS and CHONS compounds is simplified to the equation of $\overline{\text{OS}}_C \approx 2 \times \text{O/C} - \text{H/C}$ here. The uncertainty of this assumption on the calculated $\overline{\text{OS}}_C$ for CHOS and CHONS species is further discussed in Supplementary Information. The modification of the $\overline{\text{OS}}_C$ is shown in Eq. 3.

$$\overline{\text{OS}}_C \approx 2 \times \text{O/C} - \text{H/C} - (\text{OS}_N \times \text{N} : \text{C}) \quad \text{Eq. 3}$$

Where $\text{OS}_N = 0$ if $n\text{N} = 0$; This is for CHO compounds, and CHOS and CHONS compounds for LC-Orbitrap MS analysis; $\text{OS}_N = +3$ if $n\text{O} < 3$; $\text{OS}_N = +5$ if $n\text{O} \geq 3$.

2.6 Estimation of component properties

Data from both techniques can be used to investigate the gas-to-particle partitioning behaviour of the targeted compounds in the α -pinene precursor system. Saturation concentration ($C_{i,Fp}^*$), the concentration at which 50% of a component, i , is in the vapour phase and 50% is condensed, can be estimated based on the partitioning method from particle-phase fractions (F_p) from FIGAERO-CIMS. Although the homologous series of polyethylene glycols (PEG; $(\text{H}-(\text{O}-\text{CH}_2-\text{CH}_2)_n-\text{OH})$ for $n=3$ to $n=8$) were performed to calibrate the vapour pressures in the FIGAERO-CIMS, useful polyethylene glycols (PEG; $(\text{H}-(\text{O}-\text{CH}_2-\text{CH}_2)_n-\text{OH})$ for $n=3$ to $n=8$) calibrations are not accessible for interpretation of this dataset (see Voliotis et al. (2021) for a full discussion). The thermogram method for the online measurements is therefore not accessible to this study and the partitioning method is applied here. Additionally, saturation concentration of a component, i , (C_i^*) was estimated based on the structural information independently obtained from LC-Orbitrap MS.

435 2.6.1 Partitioning Method

F_p is calculated for a given species i by equation 4, from P and G ; the total signals are corrected by the sampling volume in the particle and gas phase, respectively. Some instrumental factors can affect the gas to particle partitioning process and thus affect the F_p and C^* calculated in this way, such as the deposited mass of the component on the filter (Thornton et al., 2020), gas-particle phase equilibrium on the filter, reactions on the surface or in the bulk, or diffusion of the gas to the particle-phase surface or the bulk (Mai et al., 2015; Huang et al., 2018; Yli-Juuti et al., 2017).

$$F_{pi} = \frac{P_i}{P_i + G_i} \quad \text{Eq 4}$$

C^* can be estimated by applying partitioning theory using the equation

$$445 \quad C_{i,fp}^* = OA \times \left(\frac{1}{F_p} - 1 \right) = OA \times \frac{G}{P} \quad \text{Eq 5}$$

Where OA is the suspended particle concentration in the chamber, which is calculated based on the AMS measurement, $\mu\text{g m}^{-3}$.

2.6.2 Molecular structural method

The structure of targeted compounds is identified using the LC- Orbitrap MS, after which vapor pressure is estimated using the EVAPORATION method (Compernelle et al., 2011) in the UManSysProp tool developed by Topping et al. (Topping et al., 2016). The EVAPORATION method has been assessed as more accurate than other vapor pressure estimation methods for vapour pressure of individual organic compounds (O'meara et al., 2014), though significant uncertainties persist, particularly for multifunctional compounds. 455 The C^* is calculated using equation 4 modified by Donahue et. al (Donahue et al., 2006) based on the earlier absorptive partitioning theory (Pankow, 1994). Errors in the estimation of C^* attributable to the uncertainties of vapor pressure estimation are extensively discussed in the literature (Barley and Mcfiggans, 2010; Bilde et al., 2015), and are not further discussed here. Additional uncertainties related to the activity coefficient in mixtures of components – 460 implicit in the derivation of C^* from CIMS and explicitly assumed as unity for the LC-Orbitrap MS.

$$C_i^* = \frac{M_i 10^6 P_{L,i}^\circ}{RT} \quad \text{Eq 6}$$

Where M_i : molecular weight of species i , g/mol;

$P_{L,i}^\circ$: saturation vapor pressure of pure compound i at temperature T , Pa; here T is 298.15K;

465 R : gas constant, $8.314 \text{ m}^3 \text{ Pa K}^{-1} \text{ mol}^{-1}$.

3. Results and discussion

We used the data processing method described in Sections 2.2 and 2.3 to generate the full formula list from FIGAERO-CIMS and LC-Orbitrap MS measurements, respectively. Assigned chromatographic/mass spectral peaks are used to explore the molecular formulae and broad chemical groupings in the α -pinene photooxidation system. Common capabilities
470 of the two techniques will be illustrated in section 3.1. Additional capabilities of each technique and the strength of their combination are shown in sections 3.2, 3.3 and 3.4, respectively.

3.1 Overview of SOA elemental composition

475 Offline filters characterized by LC-Orbitrap MS are shown in Fig 2 from negative and positive ionization mode alongside the ~~final—last cycle of~~ gas and particle-phase measurements ~~(5.5 hours after lights on) by the CIMS, 5.5 hours into the photochemistry experiment.~~ In the LC-Orbitrap MS, the common compounds in all experiments account for $\sim 74\% \pm 2.5\%$ in the negative mode and $\sim 59\% \pm 8.5\%$ in the positive mode, as shown in
480 Figure 2.

General observations can be drawn from the CIMS analysis, including i) there are smaller molecules and fewer larger ones in the gas phase, ii) compounds in the CHO group dominate the particle phase, and iii) compounds with between 8 and 10 carbon atoms dominate both gas and particle phases.

485 From the CIMS particle-phase and the LC-Orbitrap MS common compounds data, the results show that i) the majority of particle-phase signals come from the compounds with 8 to 10 carbon numbers in the CHO group; ii) a small proportion ($4\% \pm 0.8\%$) of high carbon number compounds (C_{11} - C_{20}) is observed in the particle phase; iii) the CHO group is shown to account for around $76\% \pm 2.7\%$ of the total contribution in the CIMS particle-phase signal
490 and $\sim 55\% \pm 1.4\%$ of the total contribution in the LC-Orbitrap MS negative mode; iv) the CIMS particle-phase fractional signal contribution shows high similarity for the C_8 - C_{10} and

C₁₁-C₁₅ range to the LC-Orbitrap MS negative mode; v) LC-Orbitrap MS negative and positive mode measure a higher fractional signal from compounds with carbon number greater than 16 than the CIMS particle phase, most likely as a result of the higher mass resolution; vi) in contrast, more small molecules with low carbon numbers (C₂-C₇) are observed from the CIMS particle-phase measurement, which may be due to the thermal decomposition of low-volatility molecules into smaller fragments (Lopez-Hilfiker et al., 2015; Isaacman-Vanwertz et al., 2016; Thornton et al., 2020). The fractional particle-phase contributions to the CHO and CHON groupings were comparable between the techniques, though a higher fraction of larger molecules was observed particularly in the CHON fraction in the LC-MS negative mode measurement.

From the LC-Orbitrap MS common compounds alone, the results suggest that the positive ionisation mode shows that the CHO and CHON groups contribute to around 48%±7% of total sample abundance. A large fraction (~11%±2.6%) of compounds in the CHON group was observed to contain a high carbon number ($\geq C_{16}$), showing more formulae with high molecular weight in the positive mode. Signals in the CHO group in both the positive and negative modes are dominated by the C₈₋₁₀ ions.

The inconclusive compounds in all experiments from the LC-Orbitrap MS are distributed in the four groups (grey bar in Figure 2a), accounting for 26.1% ± 6.6% and 40.6% ± 6.1% in both modes. As shown in Figure 2b, 56% ± 4.8% of the inconclusive compounds are with high carbon number ($\geq C_{16}$) in the negative mode, and the compounds with nC = 8-10 and nC $\geq C_{16}$ make large fractions in the positive mode, accounting for ~ 43% ± 4.4% and 33% ± 3.4% to the total inconclusive compounds, respectively. The results suggest that the largest fractions of compounds with high carbon numbers are found in the inconclusive compounds, although the existence of those compounds may result from some uncertain factors, such as control of experimental conditions, instrumental errors and operational differences or data processing. Irrespective of the consideration of the inconclusive compounds, a higher fraction of the signal is found in compounds with high carbon numbers in the LC-Orbitrap MS measurements than in the CIMS particle measurements.

Measurements of the oxidation products from both techniques in one representative experiment can be represented in terms of three different chemical spaces as shown in Figure 3. Figure 3a shows that more compounds with higher carbon numbers (>10) are observed in

the CIMS particle phase compared to that in the gas phase. In the particle phase, Figures 3a and 3b show that the results from LC-Orbitrap MS in both ionisation modes show more molecules with higher carbon numbers ($C \geq 16$) and higher oxygen number molecules ($O \geq 15$) than the results from CIMS particle measurement, though with some of them are from inconclusive compounds. More compounds with lower carbon numbers (< 5) are mostly detected by the CIMS particle measurements; likely owing to high molecules thermally desorption into small molecules in the online particle-phase measurements or the inability to measure high volatility compounds using LC-Orbitrap MS.

To explore the chemical properties of those compounds and avoid the overlapped formulae with the same carbon number and oxygen number but different hydrogen number, such as the $C_{10}H_xO_4$, carbon number vs \overline{OS}_C and O:C vs H:C are shown in Figures 3c - 3f. Figure 3c and 3d display that some molecules with low carbon number ($C < 5$) and high \overline{OS}_C ($\overline{OS}_C > 0$) are detected by the CIMS, while molecules with $10 < nC < 20$ and $\overline{OS}_C > 0$ and molecules with high carbon number ($C > 20$) are only measured by the LC-MS techniques, although some of them are inconclusive compounds. Figure 3e and 3f show that the results from the two instruments are dominated by formulae with $1.3 < H/C < 1.8$ and $0.3 < O/C < 0.7$ which show in the center of the diagram with high density. The majority of compounds in the left bottom of the diagram with a low H: C ratio (< 1) and O: C ratio (< 0.3) are compounds in the S-containing subgroups, which are only identified by LC-Orbitrap MS. Some highly oxidized (O: C > 1) compounds are detected by both techniques.

The results suggest that challenges remain in the identification of SOA products using only one technique owing to different capabilities and preferences of instruments. For example, the electrospray ionisation used in the LC-Orbitrap MS has different ionisation efficiencies and sensitivity to molecules, i.e. more sensitive to carbonyl species in positive mode and carboxylic acid or other polar species in negative mode (Mehra et al., 2020b). The iodide-adduct ToF-CIMS is more sensitive to the polar molecules and sensitivity increases with the addition of a polar functional group in the order of keto-, hydroxy and acid groups (Lee et al., 2014). Thus, the benefits of using combinatorial analytical techniques enable different insights into components found in SOA particles.

3.2 Temporal change in elemental composition

The evolution of gas- and particle-phase oxidation products grouped according to their elemental formulae and carbon numbers is shown in Fig 4. Figure 4a shows that the gas-phase products are initially dominated by CHON species at the beginning, with a decreasing trend in relative contribution over time. In Figure 4b, CHO species are shown to dominate in the particle phase, increasing over time in relative contribution, plateauing after 2.5-h into the experiment. ~~displays that, while CHO species are dominant in the particle phase, increasing over time in relative contribution, plateauing after 2.5-h into photooxidation.~~ The NO₂/NO ratio increases to a maximum value after 1 hour and then decreases (Figure S2a), suggesting the CHON species are mostly generated at the beginning with high NO conditions. In the gas phase, the C₈₋₁₀ ions dominate the CHO and CHON groups. The fraction of C₈₋₁₀ ions in the CHO group increases over time, whereas the fraction of C₈₋₁₀ ions in the CHON group decreases with time. For the particulate oxidation products (Fig 4b), there is a rapid increase to a constant maximum C₈₋₁₀ CHO signal (~42%) and a smaller and slightly reducing C₈₋₁₀ CHON (~15%) over time. There is a significant early increasing signal contribution from compounds with carbon numbers greater than 15 in the CHO group. The slight increasing fraction of high carbon number compounds (C_{≥16}) is investigated in the particle phase over time. It is not possible to confirm whether compounds with high carbon numbers are generated from either gas-phase chemistry reactions or particle-phase reactions in this study.

3.3 Clustering of the oxidation products from FIGAERO-CIMS

The application of HCA to cluster the top 100 CIMS gas-phase and CIMS particle-phase oxidation products (accounting for more than 96% of the total assigned peaks' signals) is shown in Figure 5 and Figure 6, respectively. Since the data are normalized to the highest intensity of each ion, the results are not sensitive to the intensity of raw ion signals. A dendrogram tree is applied to isolate the clusters. The number of clusters is determined according to whether it already adequately describes the temporal variation and the degree of similarity between clusters. Here, five clusters are chosen for the gas phase and six clusters are chosen for the particle phase according to these criteria. Figure 5a and Figure 6a display the average time series of each cluster calculated by averaging the signals of all ions in every measurement. The time series of ions in each cluster in the gas and particle phase are shown in Figure S4 and Figure ~~S6~~-S5 respectively, and formulae information in each cluster is provided in Table S2 and Table S3 respectively. It can be seen that the ions with similar time-

series trend are mostly clustered together. The matrix showing the relative Euclidean distance
585 between each cluster pair is provided in Figure 5b and Figure 6b, showing a clear distinction
between the clusters, with low similarity and high confidence in the quality of the clusters.

In the gas-phase HCA, cluster 5 with the highest cluster contribution peaks at the beginning
of the experiments (around 1 hour) before reaching a sharp maximum followed by a more
gradual decrease until the end of the experiment. The majority of the ions in this cluster are
590 nitrogen-containing. Cluster 4 (containing $C_{10}H_{15}NO_5$ and $C_{10}H_{15}NO_8$) and cluster 5 display
similar rates to their peak values, but different rates of decrease after 1 hour. The temporal
profiles of cluster 5 and cluster 4 show similar temporal profiles to the NO_2/NO ratio in Fig
S2a. This suggests that these early-generation products are likely formed from the reaction of
the OH or O_3 initiated peroxy radical with NO rather than with RO_2 or HO_2 (Eddingsaas et al.,
595 2012a). As shown in Fig 5c and d, cluster 4 and cluster 5 have higher weighted average
oxygen numbers and weighted average carbon number and carbon oxidation state than other
clusters. The high number of oxygen atoms presents indicates these species may have
formed via isomerisation of the peroxy radical or secondary OH or O_3 chemistry.

Cluster 3 is the second to peak and is dominated by the CHO species, suggesting significant
600 termination by RO_2 with HO_2 or RO_2 at this stage. The lower nC and nO might be caused by
C-C bond scission during secondary photooxidation reactions over time (e.g. $C_9H_{14}O_5$,
 $C_8H_{12}O_5$, etc), while the higher \overline{OS}_C in cluster 3 is likely caused by the dominant non-
nitrogen-containing compounds (or by the fragmentation leading to a lower carbon number
whilst retaining oxygen). Cluster 2 is the third cluster to peak and increase over time,
605 consistent with a later generation of products (Eddingsaas et al., 2012b). The weighted
average nC, nO and \overline{OS}_C values exhibit the lowest nC and nO, as shown in Fig 5c and d.
Some carbon-hydrogen species (CHO) appear in this cluster, as shown in Table S2, likely
produced from the reactions of RO_2 with HO_2 or RO_2 . Some nitrogen-containing ions
(CHON) with $nC \leq 8$ are in this cluster, suggesting they have undergone a higher degree of
610 fragmentation. Cluster 1 shows the slowest formation rate and lowest cluster contribution
suggesting they are likely later generation products.

Particle-phase HCA results are shown in Figure 6, and the formulae in each cluster are listed
in Table S3. Cluster 5 exhibits the highest contribution to the total identified peaks signals. It
shows a continuous increase and slight decrease at the end of experiments, suggesting its
615 constituent ions continuously increase in concentration in the particle phase. The majority of

ions in cluster 5 are non-nitrogen-containing with 7 or 10 carbon atoms, and which may partition from the gas phase directly (Zhang et al., 2015). Several dimers (e.g., ion $C_{19}H_{28}O_7$, ion $C_{18}H_{26}O_7$, ion $C_{17}H_{26}O_7$, etc) contribute to cluster 5, which is expected from further accretion reactions (Kristensen et al., 2014; Zhang et al., 2015). The similar temporal profiles
620 between cluster 5 and SOA mass (Figure S2b) indicate cluster 5 contains the most species with SOA mass. Cluster 4 has the second-highest contribution to the total identified peaks signals. The majority of the ions in this cluster are C_{9-10} , suggesting less C-C bond cleavage happened in this cluster. The remaining clusters (cluster 1-3 and cluster 4) show sufficiently different temporal profiles to warrant distinct clusters, but each contains few ions and their
625 cluster contributions to the total signals of identified peaks are lower than 5%.

Many of the top 100 gas-phase oxidation products are also found in the top 100 particle-phase ions. Some ions in the same gas-phase HCA cluster may be distributed into different clusters in the particle phase. For example, ions $C_{10}H_{16}O_3$ and $C_{10}H_{16}O_4$ in the gas-phase cluster 5 are in the particle-phase cluster 2 and cluster 3, respectively. Different clustering
630 between gas- and particle phases indicate that the particle components time-series trend does not unambiguously relate to the time trend of the corresponding gas-phase ion. Additionally, some ions with high carbon numbers ($nC \geq 16$, e.g., ion $C_{19}H_{28}O_7$, ion $C_{18}H_{26}O_7$, ion $C_{17}H_{26}O_7$, etc) in the particle phase are not included in the top 100 gas-phase ions.

Whilst a common peak-list is used for CIMS data, the HCA is conducted for individual
635 experiments and a single representative experiment is presented here. The effects on clustering of the variability of concentrations and temporal evolution are not straightforward to understand and present as uncertainty that allows ready interpretation. More than 85 ions are the same in the top 100 for the repeat experiment in both gas and particle phases, accounting for 92% - 99% of the top 100 ions. Gas and particle-phase HCA results for the
640 repeat experiment are provided in Figure [S7S6- S10-S9](#) and Table S4-S5. The generated cluster name from the *MATLAB* code differs between experiments (e.g., gas-phase cluster 4 in the representative experiment corresponding to cluster 3 in the repeat experiment), even though their time-series trends and chemical properties are highly similar. In order to easily compare the two experiments, the time-series trends and clusters' fraction in the
645 representative experiment were considered as the benchmark to adjust the cluster name in the repeat experiment for the gas and particle phase, respectively. The cluster fraction was taken as the particle-phase HCA reference since only six values (peak areas from six particle-phase

desorptions) were used to do the particle-phase HCA and the particle-phase temporal profiles will therefore exhibit greater uncertainties than those in the gas phase owing to their lower temporal resolution.

Comparing Figure [S7-S6](#) to Figure 5, it can be seen in the gas phase, cluster 3 in the repeat experiment contains fewer ions than that in the representative experiment, while cluster 1 in the repeat experiment contains more ions. Figures [S7e-S6c](#) and d show the different average nC (9.23 vs 8.55) but comparable nO (5.54 vs 5.31) and \overline{OS}_C (-0.12 vs -0.13) in cluster 3 and the different average nC (6.87 vs 7.18) but comparable nO (3.57 vs 3.34) and \overline{OS}_C (-0.03 vs -0.1) in cluster 1 for the repeat and representative experiments, respectively. In the particle phase (see in Figure [S9S8](#)), the chemical characteristics (\overline{nC} , \overline{nO} and \overline{OS}_C) are comparable for clusters with the highest and second-highest contributions between the two experiments, while the time-series profiles for the second-highest contribution clusters are slightly different, decreasing after 1 hour and after 2.5 hours, respectively. The different clustering results for some ions with low normalized signal contributions between the two experiments may indicate that those ions are likely sensitive to the experimental conditions (e.g., injected α -pinene concentration or seed concentrations), leading to different time-series trends in different experiments. The results suggest that the HCA is a powerful approach to separate ion's time-series behaviour. The chemical characteristics for clusters between the two experiments are comparable, although some individual ions differ in the time-series trend between the two experiments, resulting in different numbers of compounds in clusters. The differences do not affect the conclusions from the representative experiment due to the low contribution of those ions.

3.4 Major contributing ions and their properties

3.4.1 Isomeric contributions and evolution in gas and particle phase

The representative oxidation products with relatively high contributions (normalised fractional signal $> 0.5\%$) from the above HCA particle-phase results were selected to investigate their isomeric contributions, evolution and saturation concentrations in order to elucidate the mechanisms of their potential partitioning to the particle phase during SOA formation from α -pinene photochemistry.

A specific capability of the LC-Orbitrap MS is the ability to separate isomers and quantify their contributions, not possible with the current CIMS technique. Figure 7a shows all

680 molecular formulae with a relative contribution more than 1%, demonstrating the separation
of isomers with different retention times by LC-Orbitrap MS. Commonly, one isomer
dominates the relative contribution at a particular mass, most likely due to the limited
formation routes when a single VOC is oxidised. For example, the contribution of formula
 $C_9H_{14}O_4$ at RT= 5.977 min is much higher than the isomer at RT= 13.126 min (21.6% vs
0.5%). In the α -pinene system, the dominant contribution of only one isomer in one m/z in
685 offline measurement can support the FIGAERO-CIMS molecular assignment, although the
isomers cannot be separated by the FIGAERO-CIMS.

The FIGAERO-CIMS can semi-simultaneously quantify the evolution of the molecular
composition of compounds in the vapor and the condensed phase as shown in Fig 7b and 7c,
allowing investigation of gas-to-particle conversion processes. The partitioning behaviour of
690 these compounds was explored by investigating their time series of particle fractions, F_p , as
shown in Figure 7d and Figure [S4+S10](#). Distinct temporal profiles were observed in each case.
For example, F_p increases to maximum stable values for $C_9H_{14}O_4$ after 3.5 hours following
initiation of photochemistry, and for $C_{19}H_{28}O_7$ and $C_{17}H_{26}O_8$ after 1.5 hours, but with
 $C_8H_{12}O_4$ reducing after a rise in the initial 1.5 hours of photochemistry. As expected, the two
695 compounds in the “dimer” range ($C_{19}H_{28}O_7$ and $C_{17}H_{26}O_8$) have higher particle-phase
fractions (almost 1) than those of the “monomer” range oxidation products ($C_8H_{12}O_4$ and
 $C_9H_{14}O_4$) as shown in Figure 8.

3.4.2 Saturation concentrations

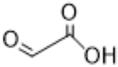
The final F_p values from the online measurements were used to calculate the C^* as shown in
700 Table 2. The LC-Orbitrap MS chromatography and mass spectrum of compounds are shown
in Figure [S4+S11](#). The structural information and estimated C_i^* based on the LC-Orbitrap
MS data are also displayed in Table 2.

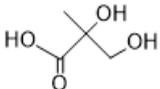
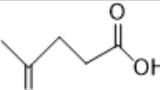
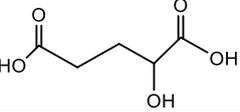
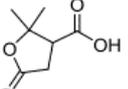
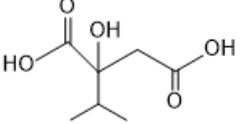
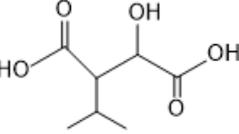
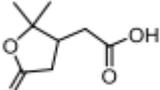
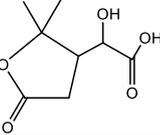
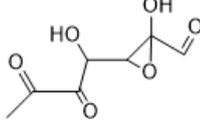
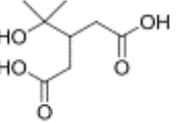
The C^* calculated from the FIGAERO-CIMS measured F_p was generally within around 2
orders of magnitude of that estimated from the LC-Orbitrap MS-derived molecular structure
705 for most compounds and sometimes much closer. Such agreement could be used to assign
classification into the somewhat broad and arbitrarily defined volatility categories (such as
semi-volatile, low volatility, intermediate volatility etc) but such utility is questionable.
Moreover, clear discrepancies were noticed for the lowest and highest volatility compounds,
as shown in Table 2. It appears that the structural assignment of $C_2H_2O_3$ and $C_3H_2O_3$ using

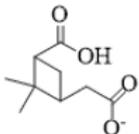
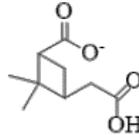
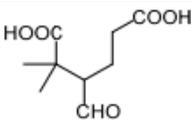
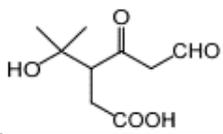
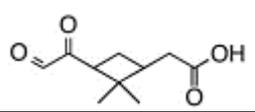
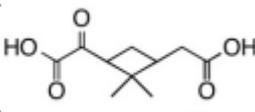
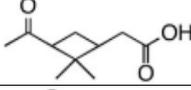
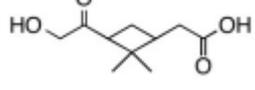
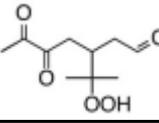
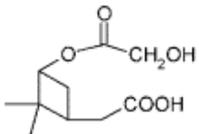
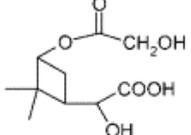
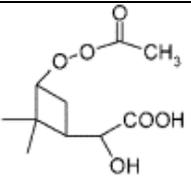
710 the MCM provides unreasonable estimates of high volatility, such that such components
 could not reasonably be expected to be found in the particle phase. The ions detected at this
 m/z in the FIGAERO-CIMS are likely from desorption fragments of larger ions as their
 desorption T_{max} are around 83°C and 105°C for C₂H₂O₃ and C₃H₂O₃, respectively. In any
 case, the physical meaning of their existence in the particle sample is questionable. It is
 715 possible that the accurate determination of the F_p for compounds with large number of carbon
 atoms (C₁₇H₂₆O₈ and C₁₉H₂₈O₇) is limited by the detection limit and signal to noise ratio in
 the gas phase measurements (Stark et al., 2017; Lopez-Hilfiker et al., 2016b). In addition to
 the difficulties in measuring partitioning behaviour, challenges remain in its prediction from
 molecular structure, owing to the acknowledged problems with vapour pressure (and hence
 720 C*) estimation, particularly for multifunctional compounds.

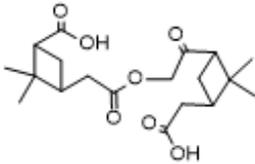
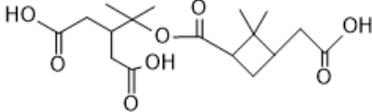
Although the C* estimation from the two instruments provides valuable cross confirmation
 for C* of the targeted compounds, the results suggest that accurate measurement for C* of
 compounds remains challenging. The differences in the estimation C* of a compound
 between the two instruments may be attributable to the different estimation methods. The
 725 structural information-based method relies on knowledge of the input of the molecular
 structure to the vapor pressure estimation model, while the partitioning method is dependent
 on accurate experimental measurements in the gas-to-particle partitioning process. For the
 partitioning approach, the measured particle fraction is likely biased owing to the thermal
 desorption of larger ions from the filter (Lopez-Hilfiker et al., 2015), or the gas-phase signals
 730 of compounds can be over measured owing to the signal-to-noise limitations (Stark et al.,
 2017; Lopez-Hilfiker et al., 2016b). The current gaps between the two methods require much
 work to further investigate the C* of compounds in the aerosol particles.

735 **Table 2.** Molecular information, being ordered in carbon number, and estimated C* of compounds
 obtained from partitioning theory method and structural information method. The uncertainty of the
 LogC* from the partitioning theory method was from the uncertainty of measured F_p and was
 propagated 1 standard deviation.

Formula	MW	RT (min)	Tentative structure and chemical name	Log C* ($\mu\text{g m}^{-3}$) ¹	Log C _{Fp} * ($\mu\text{g m}^{-3}$) ²
C ₂ H ₂ O ₃	74	n.a		6.76	2.48±0.46
C ₃ H ₂ O ₃	86	n.a		8.36	2.62±0.13

$C_4H_8O_4$	120	0.858		2.48	3.92 ± 0.05
$C_5H_8O_3$	116	2.122		4.68	5.51 ± 0.05
$C_5H_8O_5$	148	0.951		0.96	2.38 ± 0.18
$C_7H_{10}O_4$	158	2.165		5.63	3.04 ± 0.65
$C_7H_{12}O_4$	160	1.623		0.86	3.2 ± 0.48
$C_7H_{12}O_5$	176	1.398	 (Zhang et al., 2018)	1.15	2.08 ± 0.38
$C_7H_{12}O_5$	176	2.254		0.46	
$C_8H_{12}O_4$	172	2.875	 Terpenylic Acid (Claeys et al., 2009)	3.52	2.63 ± 0.42
$C_8H_{12}O_4$	172	4.536	 Norpinic acid (Winterhalter et al., 2003)	1.80	
$C_8H_{12}O_5$	188	1.326		1.26	2.09 ± 0.37
$C_8H_{12}O_5$	188	1.782		4.55	
$C_8H_{14}O_5$	190	1.94		0.47	2.58 ± 0.19

$C_9H_{14}O_4$	186	5.977		1.34	
			(Yasmeen et al., 2011)		2.4±0.32
$C_9H_{14}O_4$	186	13.126		1.34	
			carboxylic acid (Yasmeen et al., 2011)		
$C_9H_{14}O_5$	202	2.916		0.49	
					2.24±0.75
$C_9H_{14}O_5$	202	4.272		1.16	
$C_{10}H_{14}O_4$	198	2.958		1.58	2.4±0.32
$C_{10}H_{14}O_5$	214	3.432		2.21	2.11±0.28
$C_{10}H_{16}O_3$	184	7.935		3.85	3.09±0.45
$C_{10}H_{16}O_4$	200	4.297		3.91	2.57±0.41
			Hydroxy-pinonic acid (Kristensen et al., 2014)		
$C_{10}H_{16}O_5$	216	5.717		0.93	
					1.82±0.64
$C_{10}H_{16}O_5$	216	4.948		1.21	
$C_{10}H_{16}O_6$	232	3.463		-0.91	
					1.73±0.48
$C_{10}H_{16}O_6$	232	6.189		2.07	

$C_{17}H_{26}O_8$	358	12.167	 (Kristensen et al., 2014; Demarque et al., 2016)	-2.64	1.17 ± 0.38
$C_{19}H_{28}O_7$	368	14.268	 (Kristensen et al., 2014)	-0.55	1.09 ± 0.33

Note: ¹Log C* estimated from the molecular structural information-based approach. ²Log C* estimated from the partitioning theory approach.

4. Conclusions

- 740 The characterization of SOA chemical composition in the gas and particle phases is an important but complex process, containing potentially thousands of oxidation products. This study demonstrates the capabilities of the combination of FIGAERO-CIMS and LC-Orbitrap MS in the investigation of SOA chemical compositions and their properties in the α -pinene photooxidation reactions.
- 745 The distribution of particle-phase products is broadly similar between the FIGAERO-CIMS and LC-Orbitrap MS negative ionisation mode for the α -pinene SOA products. The LC-Orbitrap MS positive ionisation mode provides additional information, with more detected molecular formulae with higher carbon and oxygen numbers. Different insights into components in SOA particles can be gained from the combination of the two techniques.
- 750 The gas and particle time-series data from the FIGAERO-CIMS measurement time-series were separated by the hierarchical clustering analysis (HCA), an approach that does not rely on mass calibration. HCA was able to derive cluster properties in terms of average carbon number, oxygen number, oxidation states, for example, enabling interpretation in terms of possible termination pathways of peroxy radicals in the α -pinene photooxidation reactions.
- 755 There is substantial uncertainty in the accurate estimation of saturation concentration of compounds. C* estimation from the partitioning method using FIGAERO-CIMS measurements and molecular structure-based method from LC-orbitrap MS enables comparison of the volatility of the targeted compounds. The differences resulting from the two approaches indicates that further work is required to investigate whether the C*

760 determined for the compounds are consistent with our understanding of the processes
determining the time evolution of phase partitioning.

765 ~~There are additional questions that remain to be addressed in future work. First, instrumental
sensitivity needs more effort in order to have a better understanding of SOA chemical
composition. Second, future work will utilize the analytical approaches from the two
instruments to explore SOA oxidation products and their chemical properties from complex
systems, such as anthropogenic VOCs or biogenic and anthropogenic mixed VOCs, in the
chamber experiments. As this study only assessed a single a-pinene photooxidation system,
more work is needed to evaluate how those analytical methods from the two instruments
perform with other systems.~~

770 The selectivity and instrumental sensitivity make it impossible to use one instrument to
capture all oxidation products, thus it strongly recommends the combination of instruments in
order to better explore the evolution of SOA chemical composition. Additionally, future work
will report SOA oxidation products and their chemical properties from more complex
systems utilizing the analytical approaches from the two instruments, such as the chamber
775 photo-oxidation of anthropogenic VOCs or biogenic and anthropogenic mixtures of VOCs
(Voliotis et al., 2022). As this study only assessed a single a-pinene photooxidation system,
more work is needed to evaluate how those analytical methods from the two instruments
perform with other systems.

Data availability

780 The dataset of this study is available on the open dataset of EUROCHAMP programme
(<https://data.eurochamp.org/data-access/chamber-experiments/>).

Author contributions

785 GM, MRA, MD, AV, YW and YS conceived the study. MD, AV, YW, and YS
conducted the experiments. TJB provided on-site help deploying the FIGAERO-CIMS. KLP
and JFH offered help on the LC-orbitrap MS analysis. MD conducted the data analysis and
wrote the manuscript with inputs from all co-authors.

Competing interests

The authors declare that they have no conflict of interest.

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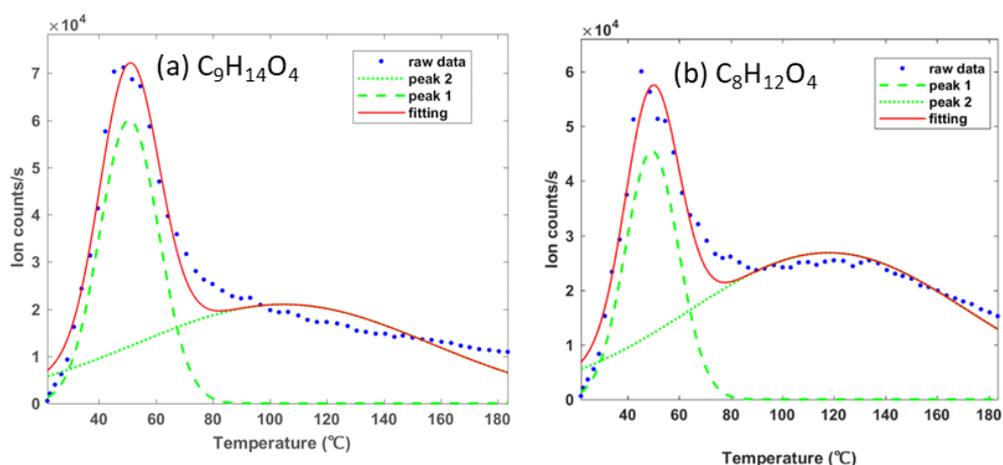
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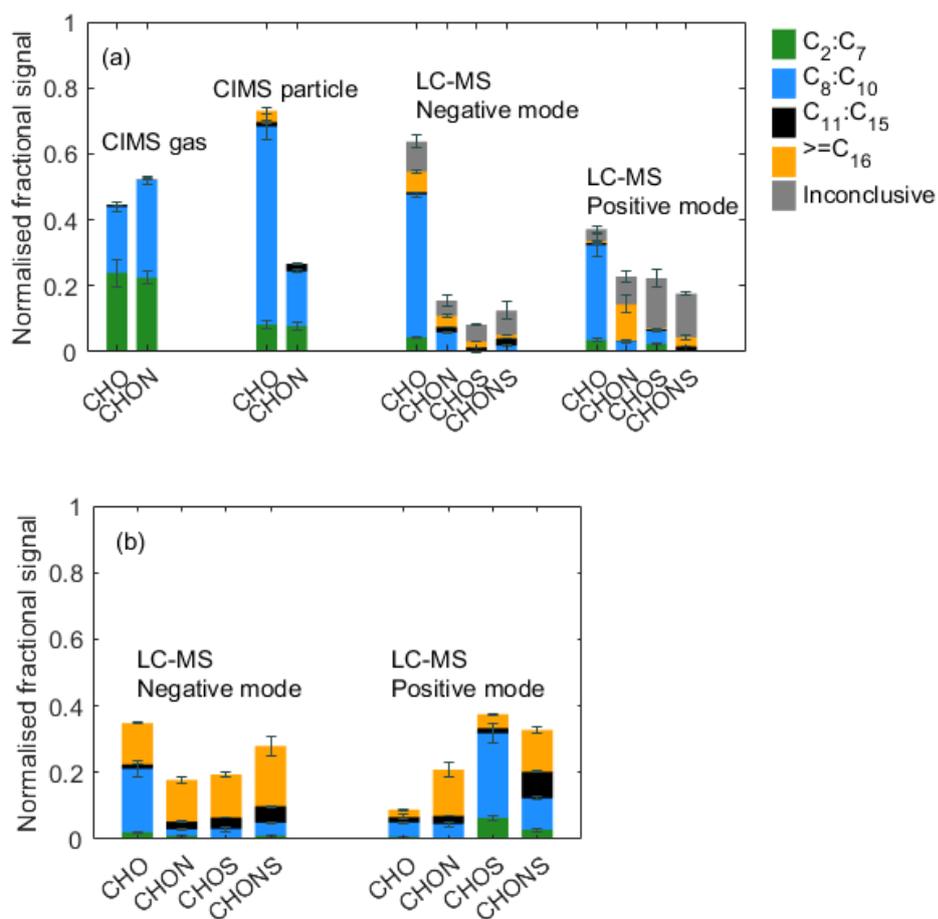
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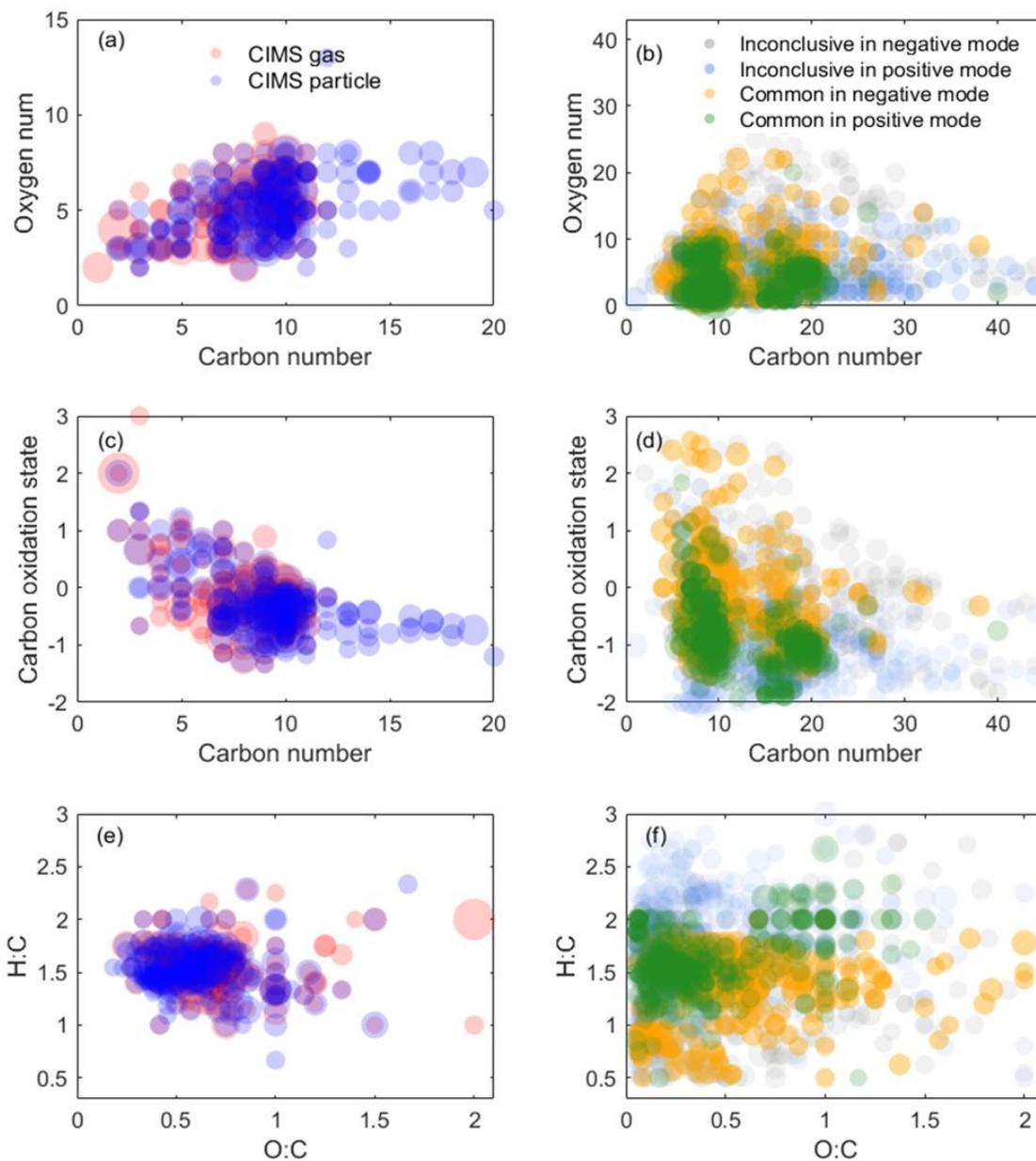
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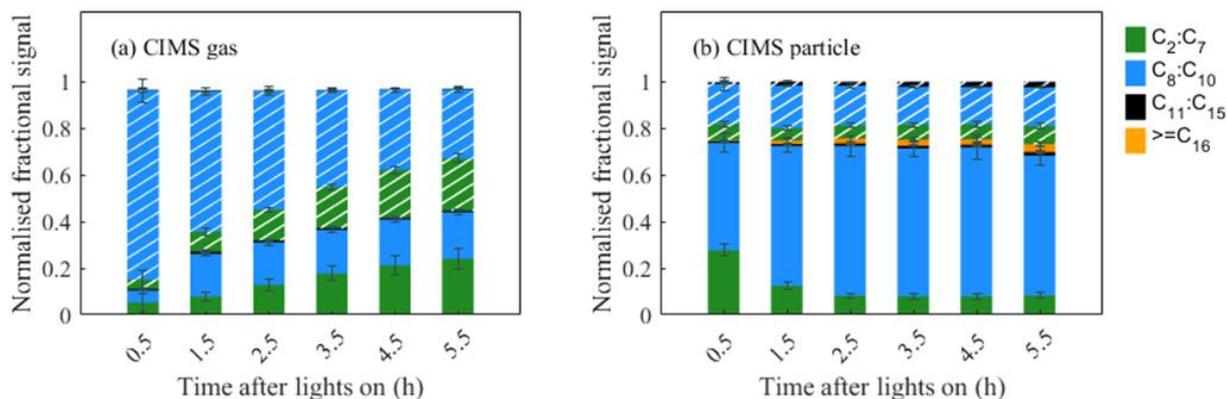
1190 Figure 1. Two examples of an ion with more than a single desorption peak in the FIGAERO-CIMS. Here the peaks at the lower desorption temperature were assigned to monomer $C_9H_{14}O_4$ (a) and $C_8H_{12}O_4$ (b) generated from α -pinene directly and the broader higher temperature peaks being a fragment of another ion at the same mass.



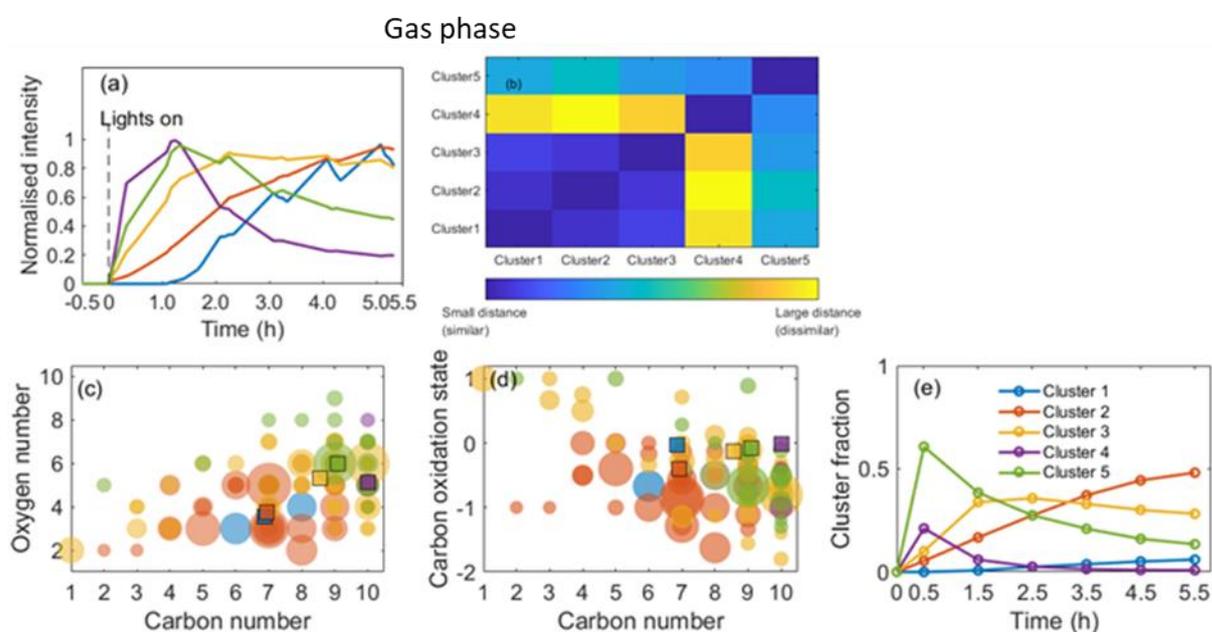
1195 Figure 2. (a) Elemental grouping from CIMS gas phase measurement, CIMS particle-phase measurement, LC-Orbitrap MS in the α -pinene system. The signal of each compound is normalized to all compounds. (b) Elemental grouping for LC-Orbitrap MS inconclusive compounds. The signal of each compound is normalized to the total inconclusive compounds. The colors correspond to the ones in (a). Standard deviations ($n=2$ for CIMS and $n=3$ for LC-Orbitrap MS) are indicated by error bars



1200 Figure 3. Oxidation products distribution from the representative experiment detected by the CIMS measurement and LC-Orbitrap MS negative and positive mode. (a), (c) and (e) are for the CIMS measurements. (b), (d) and (f) are for LC-orbitrap MS measurement. The symbol size is proportional to the square root of the ion's signal intensity.

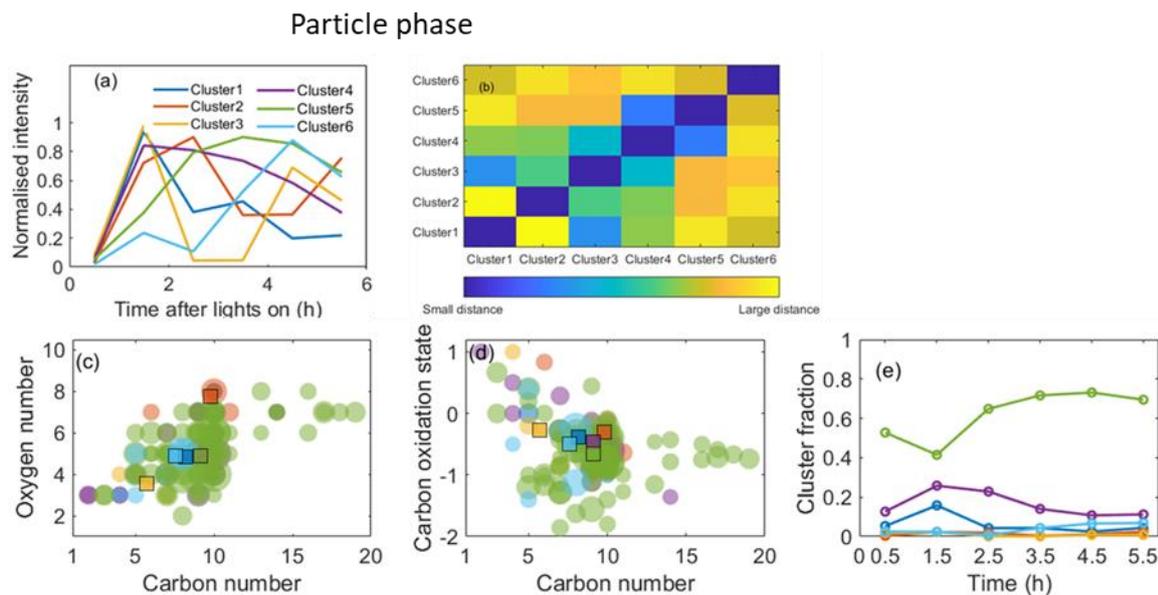


1205 Figure 4. Evolution of chemical compounds in the gas phase and particle phase. Unhatched bars are the CHO group species; hatched bars are the CHON group species. (a) CIMS gas phase; (b) CIMS particle phase. Standard deviations ($n=2$) are indicated by error bars.



1210 Figure 5. Hierarchical clustering of gas-phase oxidation products from the representative α -pinene system. (a) Time series of each cluster normalized to the highest ions' intensity between 0 and 1. (b) Matrix showing the relative distance between clusters. (c) Carbon number vs oxygen number for each cluster. (d) Carbon number vs oxidation state for each cluster. (e) Time series of the sum of ions' normalized signal fractions to the total signal in each cluster. Note that the square symbols represent the contribution weighted average carbon numbers, oxygen numbers or and \overline{OS}_C in each cluster. The colors correspond to the ones in (a).

1215



1220 Figure 6. Hierarchical clustering of particle-phase oxidation products for the representative α -pinene
 system. (a) Average time series of each cluster normalized to the highest ions' intensity between 0 and 1;
 (b) Matrix showing the relative distance between clusters. (c) Carbon number vs oxygen number for
 1225 each cluster. (d) Carbon number vs oxidation state for each cluster. (e) Time series of the sum of ions'
 normalized signal fractions to the total signal in each cluster. Note that the square symbols represent the
 contribution weighted average carbon numbers, oxygen numbers or and \overline{OS}_C in each cluster. The colors
 correspond to the ones in (a).

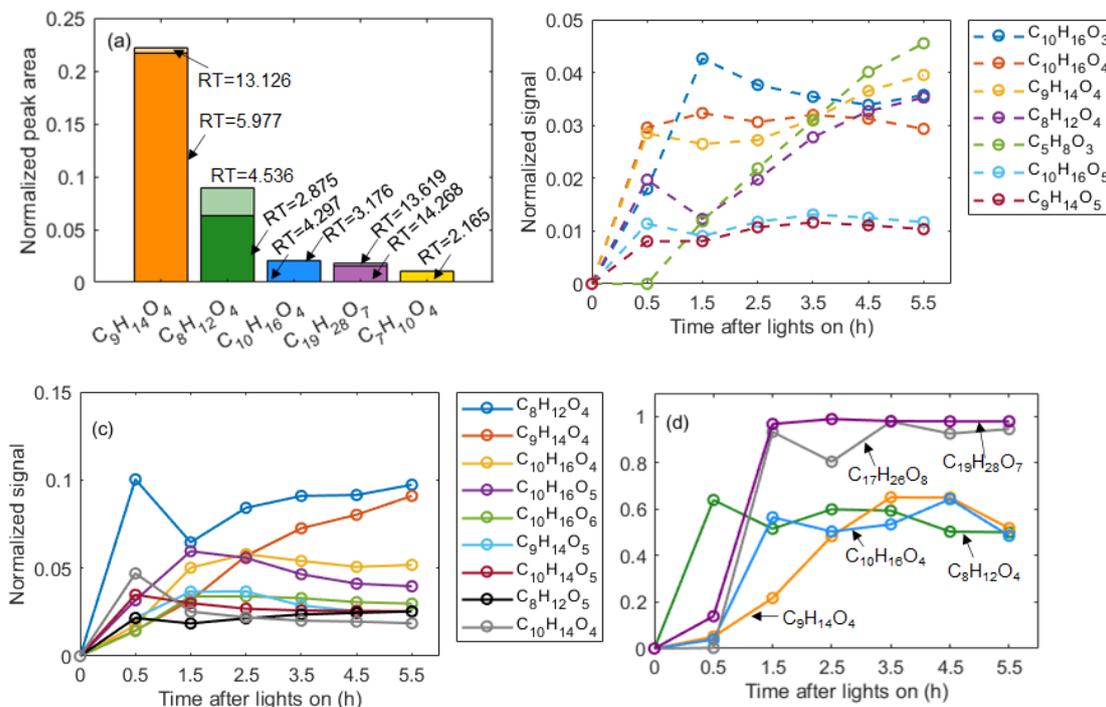


Figure 7. (a) Isomeric fractions of ions from LC-Orbitrap MS analysis; (b) Time series of compounds in
 the gas phase from CIMS gas-phase measurements. (c) Time series of compounds in the particle phase

1230 from CIMS particle-phase measurements. (d) Temporal profiles of particle fractions for a few identified organic molecules generated from α -pinene photooxidation reactions.