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A two-dimensional volatility basis set – Part 2: Diagnostics of organic-aerosol evolution

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Abstract. We discuss the use of a two-dimensional volatility-oxidation space (2-D-VBS) to describe organic-aerosol chemical evolution. The space is built around two coordinates, volatility and the degree of oxidation, both of which can be constrained observationally or specified for known molecules. Earlier work presented the thermodynamics of organics forming the foundation of this 2-D-VBS, allowing us to define the average composition (C, H, and O) of organics, including organic aerosol (OA) based on volatility and oxidation state. Here we discuss how we can analyze experimental data, using the 2-D-VBS to gain fundamental insight into organic-aerosol chemistry. We first present a well-understood “traditional” secondary organic aerosol (SOA) system – SOA from α-pinene + ozone, and then turn to two examples of “non-traditional” SOA formation – SOA from wood smoke and dilute diesel-engine emissions. Finally, we discuss the broader implications of this analysis.

1 Introduction

Organic aerosol (OA) comprises a complex system linking emissions, chemical transformations, phase partitioning and removal (Kroll and Seinfeld, 2008; Donahue et al., 2009; Hallquist et al., 2009). Until recently, OA was classified as either primary (POA) – compounds emitted directly in the condensed phase – or secondary (SOA) – condensed-phase compounds formed from volatile precursors. Those distinctions are fading as it becomes clear that the large majority of all organic compounds associated with OA are semi volatile (Donahue et al., 2006; Robinson et al., 2007), and that extensive oxidation chemistry pervades the system, resulting in high concentrations of oxidized OA (OOA) (Jimenez et al., 2009; Kroll et al., 2011).

When a gas-phase organic precursor is oxidized in the atmosphere it can generate reaction products with added functionality (oxygen containing substituents), higher polarity, and thus lowered vapor pressure (Pandis et al., 1992). It is widely accepted that a portion of these products partition into aerosol particles, with partial pressures controlled by the thermodynamics of the complex organic mixtures in the particles (Pankow, 1994; Pankow et al., 2001; Pankow and Asher, 2008). Historically, attention focused on relatively light, high-volatility SOA precursors, which dominate organic carbon mass emissions (Fraser et al., 1997; Griffin et al., 1999). However, much heavier, low-volatility vapors can potentially form SOA much more efficiently than the more abundant but much more volatile traditional SOA precursors (Robinson et al., 2007; Shrivastava et al., 2008; Donahue et al., 2009).

Field observations during the past decade provided consistent evidence that state-of-the-art models implementing standard SOA chemistry underpredict concentrations of ambient OOA in almost all environments, and often overpredict fresh POA concentrations. The magnitude of the OOA underprediction is debated, ranging from a factor of 10 (Volkamer et al., 2006; Goldstein and Galbally, 2007) to a factor of two (Gaydos et al., 2007), but it is clearly significant.

Furthermore, model shortfalls are not just OA levels but the degree of oxidation; models tend to favor relatively unoxidized POA (Chung and Seinfeld, 2002; Kanakidou et al., 2005) while observations indicate predominance of highly oxidized OOA (Zhang et al., 2005, 2007; Ng et al., 2010; Hildebrandt et al., 2010). Recent observations from high-resolution mass spectrometers give measurements of the
oxygen-to-carbon ratio (O:C) and hydrogen-to-carbon ratio (H:C) of this OA (Aiken et al., 2008; Ng et al., 2010), and these data can be combined to describe the mean oxidation state of carbon ($\overline{\Delta S_{OC}}$) of the organic aerosol (Kroll et al., 2011). Not only is ambient OA on average much more oxidized than POA, it is often more oxidized than most SOA produced in smog-chamber experiments (Jimenez et al., 2009; Holzinger et al., 2010). Parallel field observations coupling the AMS to a long residence-time (15 s) thermodenuder also reveal that both the primary and secondary components of ambient OA are relatively volatile, with half of the most oxidized organic material evaporating at 125 °C, and half of other components, including less oxidized secondary components and more reduced primary components, evaporating at 80 °C (Huffman et al., 2009; Cappa and Jimenez, 2010; Lee et al., 2011).

A challenge to our understanding of OA is that organic compounds – especially large ones – are very reactive in the atmosphere. Unsaturated compounds (mostly terpenoids including isoprene) react within an hour or two, but even large saturated organics have gas-phase lifetimes against OH reaction of five to eight hours under typical conditions, meaning that one or two generations of oxidation will typically occur in a day (Donahue et al., 2005). OH and other radicals also collide with particles, heterogeneously oxidizing the condensed-phase compounds; because of mass-transfer limitations the timescale is longer, but it is still only a few days (Moise and Rudich, 2001; Smith et al., 2003; Molina et al., 2004; Robinson et al., 2006; George et al., 2008; Kroll et al., 2009). Many oxidized organics are quite water soluble, and aqueous phase chemistry can also substantially alter OA properties and serve as an additional source of OA mass via cloud (Carlton et al., 2006; Ervens and Volkamer, 2010) or moist aerosol-phase (Ervens and Volkamer, 2010) processing.

Oxidation clearly influences OA climate effects, as organics move from being hydrophobic to hydrophilic as they gain oxygen (Cruz and Pandis, 1997, 1998; Huff Hartz et al., 2006), and hygroscopicity is related to oxidation (Jimenez et al., 2009). Oxidation could also probably influence OA health effects, though data are sparse on this topic.

We recently presented the thermodynamic underpinnings of a two-dimensional volatility space describing the coupled phase partitioning and oxygenation of organic aerosol (OA) (Donahue et al., 2011). That framework extends the one-dimensional volatility basis set (1-D-VBS) presented in Donahue et al. (2006), allowing us to describe both the phase partitioning and average composition (C:H:O) of compounds at various points in the space. A shortcoming of the 1-D-VBS is that compounds of similar volatility (e.g. tricosane and levoglucosan) can have very different properties and reactivities. OA is chemically dynamic, evolving continuously via multiphase oxidation reactions (Rudich et al., 2007; Hallquist et al., 2009), and consequently the shortcomings of the 1-D VBS make it difficult to describe OA evolution consistently.

In this work we shall use the 2-D volatility-oxidation space in a diagnostic mode, discussing ambient observations and laboratory experiments to infer important features about the chemical processes driving OA evolution. In a subsequent paper we shall present the prognostic 2-D-VBS and apply it to a series of case studies, constrained by laboratory data, to explore the role of photochemical oxidation on organic aerosol properties.

Recent work has also shown that while submicron aerosol particles “should” equilibrate quickly (Marcolli et al., 2004), mass-transfer limitations may impede the approach to equilibrium. Both diffusion limitations within particles forming amorphous, viscous, glassy solids (Zobrist et al., 2008; Virtanen et al., 2010; Vaden et al., 2011) and suppression in the apparent mass accommodation coefficient (Stanier et al., 2007; Grieshop et al., 2007, 2009c; Cappa and Jimenez, 2010; Riipinen et al., 2010) have been discussed, though these could well be manifestations of the same fundamental phenomenon. Depending on the timescale of an experiment and the extent to which conditions depart from the equilibrium state, these dynamical effects can complicate the interpretation of experiments (and be important in the atmosphere). However, they do not fundamentally alter the underlying thermodynamics. In particular, there is a misperception that partitioning models assume liquid particles. They do not. Instead, the sub-cooled liquid vapor pressure is presumed to be indicative of the free energy for any amorphous mixture (Pankow, 1994). We have not observed evidence of disequilibrium during aerosol growth experiments conducted over hours in smog chambers (once chemistry is stopped, particles stop growing) and so in this discussion we shall present an equilibrium interpretation.

2 Two dimensions

Our two-dimensional space, shown in Fig. 1a, uses volatility (saturation concentration, $C^\ast$, in $\mu g m^{-3}$) as the x coordinate and the extent of oxidation as the y coordinate. In previous publications we have used the oxygen to carbon ratio (O:C) as a measure of the extent of oxidation (Jimenez et al., 2009; Donahue et al., 2011). Here we use the average oxidation state of carbon ($\overline{\Delta S_{OC}}$) as the principal y-axis (Kroll et al., 2011). As a practical matter, “oxygenation” (O:C) and “oxidation” ($\overline{\Delta S_{OC}}$) are highly correlated (Heald et al., 2010), allowing us to use them almost interchangeably. This work complements recent work by Kroll et al. (2011), who presented the $\overline{\Delta S_{OC}}$ coordinate to describe progressive oxidation chemistry, with the organic carbon number as a second dimension.

We choose $C^\ast$ and $\overline{\Delta S_{OC}}$ as two complementary coordinates because they explicitly track properties that we can observe in experiments and in the atmosphere and that we need to know to accurately model OA behavior. Oxidation state
can now be directly measured by Aerosol Mass Spectrometry even when few of the constituent species are measured or even identified individually (Canagaratna et al., 2007; Kroll et al., 2011). Volatility can be constrained with a number of techniques, with varying degrees of certainty as discussed below.

The dimensions in our 2-D space are theoretically grounded and empirically observable. The descriptive version of the 2-D space described in Donahue et al. (2011) is built on the pure-component volatility $C^o$, from which mixing properties (gas-aerosol partitioning and condensed-phase activity coefficients) are described. The diagnostic version presented here is empirical, with observed volatility $C^*$ over whatever mixture exists. Pure-component volatilities ($C^o_i$)
and volatilities in a mixture \( C_{i,s}^* \) are related via the activity coefficient: \( C_{i,s} = \gamma_{i,s} C_i^* \). This is discussed in detail in Donahue et al. (2011).

Our starting point is that we do not know molecular structures or even carbon number for OA. This is certainly true in the atmosphere but frequently so even in chamber experiments. That is in contrast to two other 2-D spaces proposed recently, which have different objectives and strengths. The carbon-number polarity grid of Pankow and Barsanti (2009) is designed to understand and predict the relationship between carbon number, polarity, and volatility, while the carbon-number \( \overline{OS}_C \) space of Kroll et al. (2011) is designed to understand the evolution of carbon number during oxidative processes in the atmosphere. While frameworks built on the specific molecular structure of organic constituents have obvious fundamental appeal, they are only lightly constrained by observations; their power is in predicting phase partitioning for mechanisms that explicitly track molecular structure. In contrast, we know the mean oxidation state of OA samples, we know the OA levels, and we can gain a very good idea of the volatility distribution. The ansatz of our framework is that we know the OA properties far better than the specific constituents, and we thus propose a space based on those properties. This space is thus almost an inverse of the other spaces. When we discretize the space into a 2-D-VBS, our hypothesis is that the resulting "bins" will contain a collection of molecules with similar properties (specifically \( C^* \) and \( \overline{OS}_C \)). We will indeed seek to identify some of the molecules in those bins, but it is the properties that define the bins.

2.1 X-axis: volatility

Our principal axis is volatility, expressed empirically as effective saturation concentration, \( C^* \), in \( \mu g m^{-3} \) on a logarithmic coordinate. \( C^* \) determines the amount of OA \( (C_{OA}) \). The phase partitioning of a compound \( i \), meaning its fraction in the condensed phase, is given by (Donahue et al., 2006, 2011):

\[
\xi_i = \left(1 + \frac{C_{i,s}^*}{C_{OA}}\right)^{-1} \quad C_{OA} = \sum_i C_i \xi_i
\]

where \( C_{i,s}^* \) is the condensed-phase mass concentration at which 50% of compound \( i \) is in the condensed phase \( (C_{i,s} = C_{OA} \text{ when } \xi_i = 0.5) \). The total concentration \( C_{OA} \) is in turn the sum of the condensed-phase concentrations of all compounds forming the condensed-phase mixture, so Eqs. (1) must be solved iteratively.

The temperature dependence of \( C_{i,s}^* \) is given by the Clausius-Clapeyron equation, which we can approximate as (Donahue et al., 2006; Epstein et al., 2010):

\[
\ln C_{i,s}^*(T) \simeq \ln C_{i,s}^*(300K) + \left[\frac{\Delta H_{i,vap}^*}{R} \left(\frac{1}{300K} - \frac{1}{T}\right)\right]
\]

For lumped approaches such as the VBS, \( i \) is an index for a discrete bin in the 1-D or 2-D space.

Potentially as important as these equilibrium expressions, the mass-transfer rate of a compound \( i \) from the condensed phase to the gas phase will be governed by \( C_{i,s}^* \). A combination of \( C^*(300), \Delta H_{i,vap}^* \) and the accommodation (evaporation) coefficient plays a controlling role in thermodenuders (Riipinen et al., 2010; Cappa and Jimenez, 2010). Mass transfer may also be limited by diffusion in the particle phase, as there is growing evidence that the highly viscous compounds typical of OA may form a glassy condensed phase (Zobrist et al., 2008; Virtanen et al., 2010). For example, the most volatile constituents typical of reduced and more oxidized OA are pentacosane and levoglucosan, both of which are self evidently highly viscous as pure compounds (Donahue et al., 2011). This may explain the need to invoke mass accommodation coefficients far below unity in some applications (Grieshop et al., 2007; Riipinen et al., 2010). However, a glassy but amorphous condensed phase in no way invalidates the fundamental basis behind the equilibrium partitioning theory presented in Eq. (1); it merely adds the (substantial) complication that the approach to equilibrium may be delayed.

2.1.1 Measuring \( C^* \)

Volatility cannot be measured directly for relevant systems, but it is much more readily constrained than other quantities such as carbon number. The \( C^* \) distribution can be constrained from SOA growth (Pathak et al., 2007; Stanier et al., 2008), emissions dilution measurements (Shrivastava et al., 2008), volatility-based chromatography (Williams et al., 2010; Isaacman et al., 2011) and thermodenuder measurements (Grieshop et al., 2009c). All of these approaches have their challenges, including decoupling the chemistry from partitioning during SOA growth (Kroll and Seinfeld, 2008; Hildebrandt et al., 2009), uncertainties in mass transfer coefficients for both dilution samplers and thermodenuders (Riipinen et al., 2010; Cappa and Jimenez, 2010), and uncertain vaporization enthalpies for thermodenuders (Epstein et al., 2010; Cappa and Jimenez, 2010). A combination of these methods, however, can provide self-consistent descriptions of aerosol volatility distributions (Grieshop et al., 2007, 2009c).

The most direct method to measure \( C^* \) from an ambient sample is volatility-based chromatography, but the relatively polar, low volatility compounds characteristic of SOA are extremely difficult to elute (Williams et al., 2010; Isaacman et al., 2011). Absent direct measurement, \( C^* \) must be inferred.

Volatility can be inferred indirectly by driving OA out of equilibrium (e.g. changes in gas-phase concentrations or \( T \)) and observing OA responses. In almost all cases presented here, those responses are particles either growing or shrinking, it is preferable to avoid heating or cooling.
which introduces the uncertain $\Delta H_{\text{vap}}$. There are two approaches. The simplest conceptually is isothermal dilution. If a sample is diluted from, say $C_{\text{OA}} = 10^4 \mu \text{g m}^{-3}$ by a factor of 1000, the constituents with $10 \leq C_i^* \leq 10^4$ will evaporate significantly, and particles will shrink (resulting in $C_{\text{OA}} < 10 \mu \text{g m}^{-3}$). Taken as a continuous function of $C_{\text{OA}}$, size or mass data can readily be inverted for a $C^*$ distribution in volatility bins separated by a factor of 10 each (the VBS) (Donahue et al., 2006; Shrivastava et al., 2008; Grieshop et al., 2009c; Robinson et al., 2010).

The same principle applies to building up mass concentrations, including during SOA formation experiments (Odum et al., 1996; Presto and Donahue, 2006). However, for the interpretation to be straightforward it is essential that the overall (gas and particle) composition remain constant during an experiment – a highly doubtful assumption in experiments involving ongoing photochemical oxidation. Put simply, to derive volatility estimates from increasing mass yields observed following increasing precursor consumption the dominant cause must be increased formation of identical reaction products. For this reason, we have combined SOA formation and dilution experiments for $\alpha$-pinene + ozone SOA to demonstrate that the SOA volatility distributions are consistent upon formation (Presto and Donahue, 2006; Pathak et al., 2007; Shilling et al., 2008) and dilution (Grieshop et al., 2007).

Dilution is less useful as a diagnostic under most ambient conditions. This is because several orders of magnitude of dilution are required, which would eliminate virtually all of the signal. Consequently, heating in a thermodenuder is the other choice (An et al., 2007; Huffman et al., 2009; Lee et al., 2011). Thermodenuders are heated flow reactors based on shifting $C^*$ via Eq. (2). They are subject to uncertainties in $\Delta H_{\text{vap}}$ (Epstein et al., 2010; Cappa and Jimenez, 2010) as well as evaporation kinetics (Riipinen et al., 2010). For example, two recent publications using long residence time thermodenuders agree qualitatively but not quantitatively when sampling $\alpha$-pinene + ozone SOA in a smog chamber (An et al., 2007; Huffman et al., 2009). An additional issue is whether heating in thermodenuders promotes reactions in the OA. However, recognizing these uncertainties thermodenuder data can be inverted to obtain a volatility distribution and thus to place samples in our volatility space (Cappa and Jimenez, 2010; Lee et al., 2011).

2.2 Y-axis: oxidation

Our second axis is the mean oxidation state of carbon: $\overline{OS}_C$. In contrast to the principal axis, $\overline{OS}_C$ can be directly constrained even for bulk, unspcated aerosol. Specifically, it is possible to measure the important atomic molar ratios O:C and H:C, and infer $\overline{OS}_C \simeq 2O:C - H:C$ (Kroll et al., 2011).

We prefer $\overline{OS}_C$ rather than either O:C or H:C individually (O:C has been discussed previously by Jimenez et al., 2009) because oxidation reactions will invariably lead to increasing $\overline{OS}_C$. As a consequence, we expect trajectories in our 2-D space describing evolution in the atmosphere to have a monotonic increase in $\overline{OS}_C$. The formal definition of $\overline{OS}_C$ applies to the mean oxidation state of carbon atoms in a single molecule (Kroll et al., 2011), but the definition extends to the bulk aerosol. The only difference is that for a single molecule there is an exact, single value of $\overline{OS}_C$ (the average refers to the average oxidation state of carbon atoms within a single molecule), while for a bulk sample (on a single particle or an ensemble, but over many different molecules) there will be an average $\overline{OS}_C$ (technically $\overline{\overline{OS}_C}$) but also a distribution of values about that average. This is not much different than $C^*$, which has a single value for individual molecules and a distribution for mixtures.

2.2.1 Measuring $\overline{OS}_C$

The Aerodyne AMS now uses a high-resolution time-of-flight (HR-ToF) mass spectrometer with a mass resolution of about 5000 (Canagaratna et al., 2007), which is sufficient to directly measure O:C and H:C of OA samples. One can also estimate these ratios from unit mass resolution AMS samples (Aiken et al., 2008). A van Krevelen analysis of HR-AMS data finds a ubiquitous average relation $H:C \approx 2 - O:C$ (Heald et al., 2010), yielding an average relation $\overline{OS}_C \approx 3O:C - 2$ for OA samples. Therefore, one can estimate $\overline{OS}_C$ from unit mass resolution AMS data (with significant uncertainty), even when direct C:H:O measurements are unavailable. It is important to note that bulk measurements of $\overline{OS}_C$ (or OM/OC) are average properties, and that individual constituents can have very different values. Also, the formal $\overline{OS}_C$ is only approximated by $2O:C - H:C$ (Kroll et al., 2011), which in turn is only approximately measured by even the HR-AMS (Aiken et al., 2008). However, the empirical version of the 2-D-VBS permits us to define the y-axis as $2O:C - H:C$ as measured by the AMS but we can simultaneously plot individual molecules with their formal $\overline{OS}_C$ when those are known.

2.3 Dimensions and properties

An important hypothesis behind the 2-D-VBS is that important aerosol properties can be described by these two dimensions. Volatility is a given – it is one of the coordinates. The relationship between specific molecular $C^o$ and the largely empirical $C^*$ is inexact but nonetheless useful, especially given the wide dynamic range of the space. If two dimensions are sufficient, then at a minimum the average relationship between oxidation and oxygenation ($\overline{OS}_C$ and O:C) discussed above must also generally hold true – that addresses general bulk composition. Because of this, we shall generally show $\overline{OS}_C$ with O:C on a secondary y-axis in subsequent 2-D plots.

Any given point in this 2-D space represents many, many molecules. However, our conjecture is that
important properties of these molecules, such as carbon number, hydrophilicity, dipole moment, and reactivity, are similar enough that their average behaves meaningfully. Consequently, the 2-D oxidation-volatility grouping is a highly useful lumping scheme for diagnosing OA behavior and modeling the atmosphere. Individual molecules can be located precisely in the space, and trajectories associated with reaction sequences are very informative about the OA life cycle. Using group-contribution relations for volatility (Pankow and Asher, 2008) we developed a descriptive version of this 2-D space in Donahue et al. (2011); this allows us to describe the average composition (C:H:O) at any point in the 2-D space. We can accommodate nitrogen by tracking parallel spaces with specific nitrogen numbers, but for simplicity we are leaving N out of this discussion (we are effectively considering low-NO3 conditions). Once we can locate molecules, even roughly, in the 2-D space, we will be able to consider trajectories in this space defined by reaction sequences (Kroll et al., 2011). Finally, the discretized version of this space remains relatively compact and thus suitable for implementation in large-scale chemical transport models. A first implementation has already been reported (Murphy et al., 2011), and full details will be described in a forthcoming manuscript.

3 Ambient organics and the 2-D VBS

Our next objective is to assess where ambient OA lies in the 2-D-VBS.

3.1 Volatility classes of organics

To simplify discussion of different classes of organics, we shall refer to a number of volatility ranges, some of which we have defined previously (Donahue et al., 2009). In most figures these ranges are identified with colored bands. Range limits are 3 \times 10^6 because volatility bins are centered on powers of 10 (i.e. 0.3–3.0, etc). In order of increasing volatility, we have:

- Extremely low volatility organic compounds (ELVOC, gray shading, C* < 3 \times 10^{-4} \mu g m^{-3}). ELVOC are almost entirely in the particle phase under any ambient conditions, but they can evaporate in a thermodenuder.

- Low volatility organic compounds (LVOC, light red shading, 3 \times 10^{-4} < C* < 0.3 \mu g m^{-3}). These are predominantly in the particle phase in the atmosphere.

- Semi-volatile organic compounds (SVOC, light green shading, 0.3 < C* < 300 \mu g m^{-3}). Depending on ambient conditions, these compounds can be in either phase but often have sizable mass fractions in both. Note that this definition of “semi-volatile” includes much lower vapor pressures than those commonly viewed as semi-volatile (Shrivastava et al., 2008): SVOCs have vapor pressures ranging from 10^{-5} to 10^{-8} torr. The difference between common perception and this aerosol-specific definition of semi-volatile can be a source of confusion.

- Intermediate volatility organic compounds (IVOC, light blue shading, 300 < C* < 3 \times 10^6 \mu g m^{-3}). Myriad compounds of quite low vapor pressure that none-the-less reside almost exclusively in the gas phase under atmospheric conditions (Robinson et al., 2007; Donahue et al., 2009). Compounds often described as “semi-volatile” in non-atmospheric contexts lie near the high end of the IVOC range.

- Volatile organic compounds (VOC, no shading, C* > 3 \times 10^6 \mu g m^{-3}). VOCs are always represented in models either explicitly or as lumped species; we include them for completeness, because they can be important SOA precursors, and because the VOC part of the 2-D space is useful for diagnostics, as we shall illustrate below.

All of the ranges encompass thousands of individual compounds, with complexity increasing as C* decreases (Goldstein and Galbally, 2007; Kroll et al., 2011), and one objective of this work is to provide a sensible classification scheme to group and understand the atmospheric transformations of these compounds.

3.2 Empirical placement of ambient organics

Placing ambient OA and emissions data in the 2-D-VBS provides substantial insight into the dynamic evolution of atmospheric organics.

3.2.1 Ambient data

OA data from ambient AMS observations tend to group into a few characteristic components, or factors (Zhang et al., 2005; Lanz et al., 2007; Ulbrich et al., 2009). These factors can be placed in this 2-D space by combining the explicit measurement of O3C (Aiken et al., 2008; Heald et al., 2010; Kroll et al., 2011) with thermodenuder-based constraints on volatility (An et al., 2007; Huffman et al., 2009; Cappa and Jimenez, 2010). While the exact number of factors and their exact appearance (mass spectra) remains a topic of research (Ng et al., 2010), the features common to most studies include a factor resembling motor-oil or diesel emissions commonly called Hydrocarbon-like Organic Aerosol (HOA), a factor associated with biomass-burning emissions (BBOA) and one or two oxidized factors, both called Oxidized Organic Aerosol (OXA), with SV-OXA appearing to be less oxidized and more volatile and LV-OXA appearing to be more oxidized and less volatile (Jimenez et al., 2009). The most notable feature in the mass spectra representing these factors is the successive importance of an alkyl fragment, C4H9+ at m/z = 57 for HOA, a carbonyl fragment, C2H3O+ m/z = 43 for SV-OOA, and a carbon dioxide cation CO2+ m/z = 44 for...
LV-OOA. BBOA tends to be intermediate between HOA and LV-OOA and can be very difficult to isolate unless a substantial fragment associated with levoglucosan is present at m/z = 60 (C₆H₁₀O₅). The principal characteristic separating these factors is thus oxidation state. The factors separate cleanly, with $\delta^{13}C$ (O:C) of $-2.0$–$-1.5$ (0–0.17) for HOA, $-1.75$–$-0.75$ (0.07–0.42) for BBOA, $-1.0$–0.0 (0.33–0.67) for SV-OOA and 0.0–1.0 (0.67–1.00) for LV-OOA (Aiken et al., 2008; Kroll et al., 2011). This tightly constrains the y-axis location of these factors in the 2-D volatility space.

Thermodenuder measurements from both Mexico City and the Los Angeles area reveal a consistent pattern of volatility in the four AMS OA factors described above (Huffman et al., 2009). Broadly, all of the factors are fairly volatile, with significant evaporation at modest temperatures and 50% evaporation between 75 and 125 °C. The highly oxidized LV-OOA factor is the most refractory, with 50% remaining at 125 °C, while the other three factors are more or less equally volatile, with 50% remaining at 75–85 °C. While this immediately indicates that much of the organic mass in all of these factors can not be located too far from the LVOC region in our 2-D space, precisely locating the factors remains challenging.

The key features of urban OA factors (Huffman et al., 2009) are that the first three factors, HOA, BBOA, and SV-OOA (OOA-2 in Huffman) all have thermograms showing the onset of evaporation immediately upon warming from ambient temperature at a rate of about 1% K⁻¹, while the LV-OOA thermogram (OOA-1 in Huffman) is initially almost flat and then shows much more gradual evaporation. Because the thermograms for the first three factors span more than 100 K, we assume that compounds constituting these factors are distributed over 4 to 5 C⁺ bins, starting at 100 µg m⁻³ or so. The LV-OOA factor appears to be even more broadly spread out.

One recent VBS model of ambient thermograms finds the bulk of material in the HOA, BBOA, and SV-OOA factors at roughly 0.01 µg m⁻³ or higher, while LV-OOA is considerably less volatile (Cappa and Jimenez, 2010). A second model treatment of highly processed OA from Crete suggests that LV-OOA may also be mostly found with $C^+ > 0.01$ µg m⁻³ (Lee et al., 2010a). Whether these differences reflect differences among thermodenuders, variations in ambient OA properties, or differences in interpretation has yet to be determined.

Figure 1a locates the AMS factors in our 2-D space. The ovals locating the factors in this figure are not meant to be limits but rather to roughly indicate a contour that includes approximately 50% of the material responsible for each factor in the AMS. We have also not yet made any effort to tilt these ovals to show any correlation between C⁺ and $\delta^{13}C$ within the factors. Figure 1a also indicates the locations of most primary emissions, which are generally quite reduced ($\delta^{13}C \leq -1$ or so) and quite volatile.

### 3.2.2 Emissions data

Figure 1b shows our best estimate of emissions into the continental United States, as discussed in Donahue et al. (2009). Here anthropogenic (blue) and biogenic (green) emissions are separated, but each is very heavily weighted toward the VOC end of the volatility range. The anthropogenic emissions are largely pyrogenic – they involve elevated temperatures – while the biogenic emissions occur near ambient temperatures. For this reason the anthropogenic emissions surpass the biogenic emissions for $C^+ \lesssim 10^3$ µg m⁻³, and there is every reason to believe that the “primary” AMS factors – HOA and BBOA – simply represent the lowest volatility tail of these primary emission distributions. We note that while the emissions in Fig. 1b include some biomass burning listed in the US national emissions inventory (NEI), there is in all probability a missing component roughly filling the range indicated by “Biomass Burning Vapors” not in the inventory. The vapors and OA for hydrocarbons and biomass burning emissions are shown as distinct loci in Fig. 1a only to emphasize that they happen to reside in different phases under ambient conditions.

### 3.2.3 Secondary organics

The “secondary” AMS factors – SV-OOA and LV-OOA – are not consistent with known primary emissions sources. They are much more oxidized than most known emissions. Even though their exact locations in Fig. 1 are uncertain, they still define a target for proposed mechanisms, and the target is not necessarily easy to reach in models. Because emissions are largely reduced and predominantly volatile, the chemistry forming SV-OOA and most notably LV-OOA must involve substantial oxidation combined with decreases in volatility.

A final constraint is shown in the upper right-hand corner of Fig. 1a; oxidation chemistry will produce highly oxidized small organics – CH₂O, CO, and ultimately CO₂, if given sufficient time (Kroll et al., 2011). Chemical reaction sequences will trace out trajectories in this space – defining those trajectories is one of the objectives of this paper – but all trajectories must eventually veer toward the upper right in our 2-D space because of the thermodynamic imperative to produce CO₂.

### 4 Secondary organic aerosol

We know roughly where the SV-OOA and LV-OOA factors fall in our 2-D space, even though the exact C, H, and O numbers are uncertain. The next question is, how are they formed and what are the important precursors? One thing we know is that oxidation must be involved. Based on the thermodynamics developed in Pankow and Asher (2008) and Donahue et al. (2011), it is likely that many OOA molecules contain 5–12 carbon atoms. If the number of carbon atoms ($n_C$) in the carbon backbone does not change during this oxidation,
products will tend to form along lines of constant $n_C$ – diagonal lines extending from the lower right to upper left in the 2-D-VBS (Donahue et al., 2011).

One way to form LV-OOA and SV-OOA containing C$_5$–C$_{12}$ oxidized organics is to start with C$_5$–C$_{12}$ precursors and simply oxidize them, moving approximately along the carbon isopleths (conserving $n_C$), starting in the lower-right VOC region and progressing diagonally upward toward the upper-left OOA region. Without question this occurs. It is traditional SOA formation (at least in the 1st generation of oxidation), and the precursors that have dominated experimental studies – isoprene, monoterpenes, alkyl-benzenes – fall exactly in this range (Kroll and Seinfeld, 2008). A second way is to start with larger precursors that then fragment, once again producing C$_5$–C$_{12}$ oxidized organics in the LV-OOA range (Robinson et al., 2007). A third way is to start with small precursors, form small, reactive OVOCs, and then create higher carbon number association products (oligomers) from those monomers (Kalberer et al., 2004; Volkamer et al., 2007).

### 4.1 SOA formation from VOC

As an example of traditional SOA formation, in Figs. 2, 3 and 4, we analyze the formation of SOA in the laboratory from the $\alpha$-pinene + ozone reaction under low-NO$_x$ conditions with butanol present to scavenge OH radicals. This reaction is a “simple” system where we assume that a uniform distribution of products is generated as the $\alpha$-pinene is oxidized at its lone double bond (Shilling et al., 2008). There are two constraints to the 2-D product distribution. One is the systematic increase in aerosol mass yields with increasing C$_{OA}$, illustrated in Fig. 2a (Presto and Donahue, 2006; Pathak et al., 2007; Shilling et al., 2008; Kroll and Seinfeld, 2008) and driven by Eq. (1). The other constraint is the systematic decrease in oxidation state with increasing C$_{OA}$. This is widely reported for terpene + ozone reactions (Zhang et al., 2006; Grieshop et al., 2007; Shilling et al., 2009), and illustrated in Fig. 2b with HR-AMS data from Shilling et al. (2009) for $\alpha$-pinene + ozone. Dilution experiments show that SOA formed from this reaction evaporates as expected, with a corresponding increase in $\overline{\text{OS}_C}$, confirming that many of the SOA products are semi-volatile (Grieshop et al., 2007).

We assume that the overall (particle + gas-phase) product distribution remains the same in these experiments. If this is so, the changes in Fig. 2 are driven by partitioning. The figures are thus best read from left to right. The very low mass yields for C$_{OA} < 1$ µg m$^{-3}$ indicate very low yields for products with $C^*$ below this value. Mass yields for $C^* \approx 1$ µg m$^{-3}$ are near 0.05; only half is in the condensed phase, so the aerosol mass yield is 0.025 (Presto and Donahue, 2006). The bulk $\overline{\text{OS}_C}$ for $C^* \approx 1$ µg m$^{-3}$ is near −0.5 (Shilling et al., 2009). As C$_{OA}$ rises, the initial SOA with high $\overline{\text{OS}_C}$, low volatility, but also quite low concentrations is overwhelmed by more abundant and more volatile compounds having progressively lower $\overline{\text{OS}_C}$, driving the bulk $\overline{\text{OS}_C}$ lower.

The overall VBS product distributions inferred from experiments are shown in Fig. 3, in two dimensions in the top panel and one dimension below. We infer the yields of more volatile products with $C^* \geq 10^4$ µg m$^{-3}$ to close the carbon mass balance. The volatility distribution of the products is based on earlier SOA mass yield fitting in Presto and Donahue (2006), augmented to conserve carbon by Donahue et al. (2009).

![Fig. 2.](image)

 Determination of $\overline{\text{OS}_C}$ is not so straightforward. The total carbon yields in each $C^*$ column is constrained, but these can be distributed in $\overline{\text{OS}_C}$ space. Again we work from left to right. We start with an average $\overline{\text{OS}_C} \simeq −0.1$ at $C^* =$...
Fig. 3. Products from the ozone + α-pinene reaction in one and two dimensions. (a) The product mass distribution is shown with blue contours. Representative 1st generation products are indicated with colored circles within the contours, along with their molecular structures. α-Pinene is shown with a red circle. The dashed red lines extending from the α-pinene point of origin indicate the trajectories for functionalization in this 2-D space. The middle line is near the value for organic acids and also the average value assumed for 2-D-VBS calculations. (b) A 1-D-VBS product distribution for the same SOA, given by vertical sums over OS_C for each log_{10} C^{*} bin. Partitioning is shown for ∼10 µg m^{-3} of SOA, obtained from oxidation of 100 µg m^{-3} of α-pinene, with condensed-phase OA in green and vapor-phase products in white.

0.1 µg m^{-3}, arbitrarily spreading the yield over two bins in OS_C. Because the data indicated a steady drop in OS_C with increasing C_OA, the carbon yields are distributed in less oxidized bins for more volatile products. While a least-squares fit could constrain the average of these OS_C values in each 2-D column, the distribution about that average is not constrained by the bulk observations. In this work we spread the products in OS_C to obtain a smooth distribution that both agreed fairly well with the bulk AMS data and also some known products.

The product distribution compares reasonably well with a recent calculation based on a master chemical mechanism simulation of the system (Chen et al., 2011). In the upper panel, the product distribution is shown by blue contours, while in the lower panel the 1-D yields (the vertical sum of the 2-D yields) are shown with green (particle) and white (gas-phase) products for C_OA ≃ 10 µg m^{-3}. The distributions in both phases are also shown as histograms in Fig. 4, with colors indicating the volatility ranges described above. The histograms are the discretized product distributions required for model implementations of this SOA process, using the
volatility basis set in either one or two dimensions. Thus the distribution in Fig. 4a is the 2-D-VBS product distribution for this reaction. A matlab file with 2-D stoichiometric (carbon) yields is provided in the Supplement.

The predictions of mass yield and $\overline{OS_C}$ vs. $C_{OA}$ for this distribution are also shown as curves in Fig. 2. While the $C^*$ (first dimension) values are constrained by the data, the $\overline{OS_C}$ distribution as a function of $C^*$ is not – only the bulk average value is. Note that the y-value ($\overline{OS_C}$) in the 2-D-VBS is the average oxidation state of carbon in a given bin, but the overall bulk average is measured by the AMS. We thus again rely on an educated guess for a reasonable distribution in this second dimension, while keeping the average in line with observations. The range in Fig. 2b is narrow, covering the middle third of the full $\overline{OS_C}$ range used in the full 2-D-VBS; consequently, the model-measurement agreement is reasonably good.

We have also plotted some observed products of the reaction with colored circles in Fig. 3a, using $C^*$ values predicted by the SIMPOL group-contribution method (Pankow and Asher, 2008). Pinonaldehyde (yellow circle) is the highest-yield and the highest volatility major product, accounting for 20–25% of the carbon mass (Yu et al., 1999; Jang and Kamens, 1999; Jaoui and Kamens, 2001; Saathoff et al., 2009), while cis-pinonic acid (magenta circle) and picinic acid (maroon circle) are lower yield (5–10% of the carbon) but much lower volatility products. A potential hydroperoxide formed after OH elimination from a Criegee intermediate is also shown as a green circle. The product distribution shown in Fig. 3b is also broadly consistent with a master chemical mechanism simulation of $\alpha$-pinene + ozone chemistry, which in turn is consistent with the mass yields shown in Fig. 2 (Chen et al., 2011). Note that, for specific molecular products the y coordinate is strictly $\overline{OS_C}$, as the average relationship between O:C and $\overline{OS_C}$ (O:C = 1/3$\overline{OS_C}$ + 2/3) does not hold for individual molecules (i.e. unsaturated and thus somewhat oxidized $\alpha$-pinene itself).

Figure 3a also shows a wedge of 2-D space one would expect products to occupy if functionalization were the only significant chemical process leading to $\alpha$-pinene products. The relations for $-\text{OH}$ and $=\text{O}$ functionalization as well as the assumed average (similar to organic acid functionalization) are shown as red dashed lines: this wedge follows the ranges described in Donahue et al. (2011) and shows that the observed (first-generation) VBS product distribution and the properties of specific product molecules all fall within the general region expected based on the functionalization wedge.

It is also of interest to note the location of the OOA factors, as well as macromolecules (“oligomers”) that are constructed from monomers in the $\alpha$-pinene SOA distribution shown in Figs. 3 and 4. The first-generation products barely reach into the SV-OOA range, overlapping for the highest ambient $C_{OA}$ cases where volatile SVOC may contribute to OA. They are far from the LV-OOA range. Dimers would fall in the range shown by the dashed oval labeled ‘oligomers’. These dimers constructed of the first-generation $\alpha$-pinene monomers miss the OOA target; however, there is evidence for their formation in laboratory experiments (Kalberer et al., 2004; Tolocka et al., 2004; Heaton et al., 2007, 2009). While these oligomers do move the volatility distribution from the reaction closer to the “target” provided by LV-OOA, because they are non-oxidative reactions they do not address the degree of oxidation seen in LV-OOA. The same is true for suppressed evaporation of semi-volatile SOA (Vaden et al., 2011); this could decrease the apparent volatility of the SOA but would not address the shortfall in oxidation. More highly oxidized monomers may be more likely to contribute to atmospheric LV-OOA, as they would start out
with a sufficiently high $\tilde{O}_{\text{SC}}$. However, we must be cautious if significant condensed-phase oligomerization occurs with variable yields over the course of the $\alpha$-pinene SOA experiments, as this would invalidate the important assumption that the overall product distribution remains constant during an experiment, and thus complicate the interpretation of Fig. 2a.

In summary, carbon mass balances of chamber experiments, AMS $\tilde{O}_{\text{SC}}$ data, dilution experiments, and the properties of known first-generation monomeric products form a self-consistent picture of “simple” SOA formation from $\alpha$-pinene + $O_3$.

### 4.2 SOA formation from SVOC and IVOC

Another important class of precursors is the lower-volatility organic vapors (IVOCs and SVOCs) shown in Fig. 1. In this section we describe the transformation of these materials and how that is related to the “targets” described in Sect. 3.

As soon as OH radicals are introduced into a system, the condition that the product composition remain constant in time is violated. Data inversion (determination of a product $C^*$ distribution from SOA mass yields) is much more difficult because the product distribution is almost guaranteed to evolve continuously during an experiment when OH is involved. These become aging experiments, with oxidation products themselves subject to further oxidation and thus a continually evolving, multi-generational mix of products. Our ultimate approach to quantifying yields will be to develop a full 2-D-VBS OA aging model and to compare forward model predictions with the chamber data. We shall present this prognostic analysis in a subsequent publication. However, we can use the 2-D space in the diagnostic mode to assess the general characteristics of aging chemistry. To illustrate this we shall now consider two important examples of aging applied to systems traditionally associated with POA – woodsmoke and diesel emissions.

There are significant atmospheric emissions of SVOCs and IVOCs (typically with carbon numbers $> C_{10}$ or so) (Donahue et al., 2009; Robinson et al., 2010), as shown in Fig. 1. Many of these vapors are co-emitted with POA from traditional sources such as motor vehicles or biomass combustion. They can also form from evaporating primary particles as they dilute in the atmosphere (Robinson et al., 2010). In recent work we have shown that vapors from these sources can be a significant SOA source (Robinson et al., 2007; Grieshop et al., 2009b; Presto et al., 2009; Miracolo et al., 2010). These compounds partition to the gas phase as the emissions are diluted in the atmosphere (Lipsky and Robinson, 2006; Shrivastava et al., 2008; Dzepina et al., 2009; Tsippouri et al., 2010), and photochemical oxidation leads to substantial SOA formation, often doubling or tripling the aerosol mass in experiments (Weitkamp et al., 2007; Grieshop et al., 2009b). There is often much more added SOA mass than traditional, volatile precursors can produce, and based on our analysis of dilution data the concentration of vapors in these experiments with $100 < C^* < 10^5 \mu g \text{ m}^{-3}$ ($C_{15}$–$C_{20}$ compounds) is a factor of 3–10 greater than the aerosol mass at the beginning of the experiments.

AMS spectra of OA from photooxidation of wood smoke (Grieshop et al., 2009a) and diesel emissions (Sage et al., 2008) provide constraints on the progressive oxidation of the SOA formed from these emissions. Unlike the well-constrained $\alpha$-pinene system illustrated in Fig. 3, evaporated primary emissions comprise a complex ensemble of precursor vapors forming condensed-phase products with a broad range of volatilities. We must thus confront both progressive oxidation chemistry and a distribution of precursor volatilities (and other properties) when diagnosing these systems. The 2-D volatility-oxidation space can aid this diagnostic analysis: we know the mean properties of OA throughout this space, and data constrain both coordinates as they evolve in experiments or in the atmosphere. We shall use wood smoke and diesel emissions to illustrate this process.

#### 4.2.1 Wood smoke

Wood smoke is a highly variable source (McMeeking et al., 2009), but most cases reported by Grieshop et al. (2009b) saw substantial SOA formation in a few hours of photochemical oxidation. Fig. 5 shows results from one experiment, including an inset showing the evolving oxidation during photochemical oxidation (derived from Q-AMS data based on the fraction of organic signal at $m/z = 44$, $f_{44}$) on the bottom panel and thermodenuder data (mass fraction remaining at 50 °C and 70 °C) on the upper panel. The OA mass increased by a factor of 2.5 during this experiment (Grieshop et al., 2009b).

The fresh BBOA AMS spectrum resembled other BBOA samples (Grieshop et al., 2009a), falling in the BBOA range of the 2-D-VBS shown in Fig. 1. Thermodenuder data showed that approximately half of the fresh BBOA mass evaporated at 50 °C, with little change in $f_{44}$, suggesting that most of the particle-phase BBOA was confined within a narrow range in volatility within roughly an order of magnitude of the critical value of $C^* = C_{\text{OA}}$ (Grieshop et al., 2009c). However, at 70 °C, $\frac{3}{4}$ of the mass evaporated and $f_{44}$ increased sharply, suggesting a tail of more oxidized, less volatile material in the fresh BBOA. We use these data to locate BBOA in Fig. 5. The brown oval marked “BBOA” spans about 1 decade of volatility. This oval is, very roughly, a contour surrounding ~50% of the BBOA mass.

The inset of Fig. 5 shows that the whole sample (POA + SOA) became dramatically more oxidized after only a few hours of photochemistry, and heating to 70 °C was required to evaporate half of the OA mass (with a corresponding sharp rise in $f_{44}$). In 2 h the OA mass doubled from 60 to 120 $\mu g \text{ m}^{-3}$, and by the end of the experiment $C_{\text{OA}}$ had nearly tripled (Grieshop et al., 2009b). We conclude that the aged OA was spread over a wider volatility range (at least 2 decades) and that it was much more oxidized.

www.atmos-chem-phys.net/12/615/2012/ Atmos. Chem. Phys., 12, 615–634, 2012
The inset presents $f_{44}$ (registered to the 2-D-VBS y-axis) for the whole OA sample, which is a mixture of unoxidized BBOA and SOA generated via photochemistry (BBSOA). While we do not know how much of the original BBOA material may have been oxidized, possibly via "semi-volatile pumping" (Miracolo et al., 2010), there was at least 2 times as much SOA as BBOA at the end of the experiment. If the BBOA were unaltered, we would expect it to substantially but not completely evaporate at 50 °C and almost completely vanish at 70 °C; thus the 70 °C mass spectrum at the end of the experiment should be mostly SOA. There are two potential solutions describing this time evolution. One is that the SOA became progressively more oxidized over the course of the experiment, while the other is that the SOA was always highly oxidized but that there was progressively more of it, making the mixture of BBOA and SOA continually rise in $\overline{O:C}$. Of course, a combination of these solutions is possible.

One way to assess the properties of the SOA is to subtract the POA (BBOA) contribution from the AMS spectrum. Levoglucosan is commonly used as a tracer for BBOA in ambient samples, both through chromatographic analysis (Simoneit et al., 2004) and using the $m/z = 60$ fragment (Lee et al., 2010b). A residual spectrum (at room temperature) obtained by subtracting out BBOA based on the levoglucosan marker fragment shows $O:C \simeq 0.625$ ($\overline{O:C} \simeq -0.125$) based on $f_{44}$ (Grieshop et al., 2009a), which is similar to the $\overline{O:C}$ at 70 °C at the end of the experiment. A complication is that levoglucosan is not the only contributor to the unit mass resolution signal at $m/z = 60$ (Grieshop et al., 2009a). We show a BBSOA region in Fig. 5 with $\overline{O:C} \simeq -0.125$ that is consistent with either solution at the end of the 6-h experiment. Again, this contour surrounds approximately half of the BBSOA material (though the y range of the 50 % contour is difficult to specify. This contour does not slope in $\overline{O:C}$ because there is not sufficient information to define that slope. However, the total spectrum at the end of the experiment should be a combination of BBSOA and BBOA with a roughly 2:1 mixing ratio for low thermodenuder temperature but should be more or less pure BBSOA at 70 °C, and that is consistent with the vertical mixing arrows shown.

Levoglucosan is shown in Fig. 5, and its location suggests some hazards. First, levoglucosan is far from the mean properties of BBOA as determined by AMS measurements, and one might want a tracer to be more representative of those properties. However, as we have emphasized, the ovals locating the AMS factors loosely represent a 50 % contour for the total factor mass; individual constituents can lie well outside of those contours, and this is a case in point. A more significant issue is that the $m/z = 60$ signal in these experiments decayed more rapidly than simple dilution by SOA could explain (Grieshop et al., 2009a). At least some of that decay could be because levoglucosan is semi volatile, though heterogeneous oxidation can surely have played a role as well (Hennigan et al., 2010). An ideal tracer for a source-specific factor might have an $\overline{O:C}$ similar to the mean of the factor but a $C^*$ toward the low-volatility end of the source distribution.
The source of the BBSOA is the next question. Modeling by Grieshop et al. (2009a) indicates that it can not be explained based on the measured decay of volatile SOA precursors (e.g. mono-terpenes, single-ring aromatics). Based on dilution data, we know that a large pool of vapors lie just to the right of the BBOA oval in Fig. 5 (Lipsky and Robinson, 2006; Shrivastava et al., 2006). Those vapors are located roughly by the dashed brown contour in the figure. With the onset of photochemistry, oxidation products will tend to move along the gray, constant carbon number isopleths as they functionalize, possibly moving toward the upper right as well as they fragment (these steps are indicated with magenta arrows). The BBSOA at the end of the experiment is consistent with compounds near C$_7$O$_4$, whereas the precursors are likely to have been C$_{11}$–C$_{13}$ with perhaps 2 oxygen atoms. Because traditional, low carbon-number VOC precursors cannot explain its production, and because we know the higher carbon-number precursors are present, we conclude that fragmentation (at least one C–C bond cleavage) has played a role in BBSOA formation during at least two generations of oxidation chemistry (the dashed-green contours indicating oxygen number should be proportional to generation number, with 1–2 oxygen atoms being added per generation).

In this experiment we can not yet tell whether the aggregate OSC shown in Fig. 5 rises continuously because the OSC for the BBSOA components is rising via aging or whether progressively more BBSOA with more or less constant composition is simply being mixed into the bulk spectrum. To answer that question we turn to a dataset where the SOA component is separated from a POA factor during the data analysis.

4.2.2 Diesel emissions

We have analyzed an evolving residual (SOA) spectrum during diesel-exhaust aging experiments (Sage et al., 2008). Oxidation of diesel emissions under pseudo-ambient conditions also forms highly oxidized OA in a matter of hours, as shown in Fig. 6. The inset in this figure shows OSC (estimated from $f_{44}$) for the residual aerosol (i.e., after subtracting off a static POA component closely resembling published HOA spectra). The data are shown as a function of time in the inset and in the 2-D space with colored circles and arrows in Fig. 6. We observed rapid growth in OSC of the SOA in these experiments (Sage et al., 2008). OSC rose rapidly at first and then more slowly over the final several hours, as shown in Fig. 6. Thermodenuder data from separate diesel oxidation experiments are similar to those shown for woodsmoke in Fig. 5, with a quick rise in the mass fraction remaining with the onset of oxidation followed by a more or less stable (but still relatively volatile) thermogram for the duration of the experiment (Miracolo et al., 2010).

Viewed in the 2-D space, the SOA loci from this experiment rise vertically as oxidation progresses, as indicated by the ovals labeled “SOA 1h” and “SOA 5h” in Fig. 6. Initially the SOA is barely distinguishable from HOA, but after only 1 h it is well within the range of the SV-OOA factor. By the end of the 5-h experiment (much like the wood-smoke in Fig. 5), it is near the high end of OSC for SV-OOA. Traditional SOA precursors such as toluene can explain only a small fraction of the SOA; Weitkamp et al. (2007) even spiked some experiments with large aliquots of toluene and observed little additional SOA formation. The SOA was thus
formed by oxidation of vapors from evaporated POA, indicated in the figure with a dashed black contour. Experiments on n-alkane analogues for these vapors (Presto et al., 2009, 2010) as well as evaporated mixtures of diesel fuel and motor oil (Miracolo et al., 2010) confirm that these HOA vapors are a potent SOA source.

There are two potential explanations for the progressive rise in \( \text{OS}_C \) shown in Fig. 6. One (which we consider less likely) is that SOA formed from oxidation of a succession of precursors whose oxidation roughly followed the gray carbon isopleths. Lower volatility (\( \sim C_{20} \)) vapors would form SOA immediately, while more volatile vapors would require multiple generations of oxidation to generate SOA (Presto et al., 2010). This would build up a mixture of progressively more oxidized SOA constituents. For example, \( C_9 - C_{11} \) precursors could be the source of SOA observed at 1 h (shown with an orange contour in Fig. 6), while smaller carbon-number precursors could be the major source for SOA observed at 5 h (purple contour). In this model, the SOA would be formed with minimal fragmentation.

The other explanation (which we consider more likely) is that the SOA itself aged over the course of the experiment, with progressive fragmentation of semi-volatile compounds driving the average carbon number from roughly 12 early in the experiment to roughly 7 after 5 h. For the first mechanism to dominate, the incremental SOA being formed in the final hours would have to be extremely oxidized in order to average out the less oxidized material formed earlier (when the aerosol mass increases were also steeper). Furthermore, the \( C_7 \) carbon number consistent with the SOA at the end of the experiment is inconsistent with calculations of SOA formation from measured VOC precursors (Weitkamp et al., 2007). Consequently, it is likely that aging and fragmentation of SOA formed early in the experiment (and almost certainly HOA) (Presto et al., 2009; Miracolo et al., 2010) was the major cause for the steadily rising \( \text{OS}_C \). This aging chemistry is likely dominated by oxidation of the gas-phase fraction of SVOC components (Miracolo et al., 2010).

If the second mechanism dominates over the first mechanism, there must be a significant amount of fragmentation in the overall process, resulting in highly-oxidized products with \( C^\ast \) in the LVOC and SVOC range and thus 7–10 carbon atoms. This fragmentation is consistent with our overall understanding of hydrocarbon oxidation – as substitution around carbon bonds increases, the probability of carbon bond fragmentation increases as well (Lim and Ziemann, 2009; Kroll et al., 2009; Chacon Madrid et al., 2010). For lighter hydrocarbons, this fragmentation leads to very volatile products such as formaldehyde, acetone, and other highly oxidized vapors. However, for very large hydrocarbons the intermediate products will have substituents situated more or less randomly along the carbon backbone. If molecules fragment at a random location along the carbon backbone, the mass distribution of the products will favor larger carbon numbers; the median carbon number will be about 75 % of the parent compound carbon number. Thus, after fragmentation, the modal carbon number of a \( C_{20} \) precursor is roughly \( C_{15} \), but the distribution is broad.

The data shown in Fig. 6 show conclusively that the oxidation state of SOA in these experiments can rise with time via oxidation chemistry. However, the same data were analyzed using factor analysis tools described by Zhang et al. (2005), revealing fixed factors more or less identical to HOA and the most oxidized SOA (the purple symbol at 5 h oxidation). A good solution emerged in which the fraction of the oxidized SOA simply rose with time as the overall aerosol became more oxidized (Sage et al., 2008). This is an essential issue with factor-analysis methods: to the extent that spectra of intermediate states (for example the orange symbol in Fig. 6) are linear combinations of two terminal spectra, there is no way to tell the difference between a 50-50 mixture of those terminal constituents and aerosol that is of uniform, intermediate composition. Factors can represent either two distinct pools of material with variable mixing ratios or two endpoints in a continuous and evolving distribution.

The average \( \text{OS}_C \) clearly increases during the 5–6 h typical of our experiments. While added oxygen is necessary for increased \( \text{OS}_C \), the carbon number is decreasing as well (Kroll et al., 2011), as the constituents cross a succession of carbon-number isopleths, moving continuously toward lower carbon numbers. The first reaction step may involve relatively little fragmentation with the addition of 2–3 oxygens to the carbon backbone, while subsequent steps involve both fragmentation and increasing oxygen content.

### 4.2.3 General SVOC and IVOC behavior

We have begun to explore this mechanism systematically using n-alkanes and oxygenated compounds spanning the SVOC and IVOC range. We do observe substantial SOA formation and – when the OA mass is kept very low – sharp increases in \( \text{OS}_C \) (Presto et al., 2009, 2010). Thermograms are consistent with significant fragmentation in later-generation products (Presto et al., 2009). However, alkanes alone do not generally produce high \( \text{OS}_C \) products under the same conditions as these diesel experiments, suggesting that other constituents in the diesel emissions may be important contributors.

Sequences at roughly constant \( C^\ast \) consisting of progressively more oxygenated compounds show decreased SOA formation with increasing oxygenation, along with gas-phase product molecules consistent with fragmentation (Chacon Madrid et al., 2010; Chacon Madrid and Donahue, 2011). All of these experiments support the general finding that fragmentation processes become more important with increasing substitution associated with increasing oxygenation during OA aging.

Most of this discussion focuses on products in the condensed phase (that we observe with the AMS); for example, the initial products with \( \text{OS}_C \approx -1.5 \) emerge quickly in the
Fig. 7. Two reaction trajectories in 2-D space. Contours are carbon number (gray) and oxygen number (dashed green) as in Fig. 5. SOA formed from VOCs (shown at C₈) moves rapidly upward in OₛC but reaches a minimum in C* only barely sufficient to form any SOA in the atmosphere. SOA formed from SVOCs and IVOCs (shown at C₁₈) moves more slowly upward in OₛC but reaches very low C* values. Both trajectories ultimately result in complete fragmentation (in the limit where the chemistry is allowed to continue to completion), and both consist of a cloud of individual reaction pathways initially centered on the mean carbon contour for the parent carbon number. A potential oligomerization trajectory is shown as well.

5 Conclusions

We have described the use of a two-dimensional volatility-oxidation space as a tool for diagnosing organic-aerosol behavior. The coordinates are organic volatility (C*) and the average carbon oxidation state (OₛC). Because we know the average composition of compounds in this space (Donahue et al., 2011), we plot observational data and gain insight into the chemistry driving OA evolution. An important feature of the diagnostic mode is that we do not need to know the molecular composition in the space to infer broad patterns in OA behavior. Data for traditional SOA formation (the α-pinene + ozone system) are self-consistent when considered in this space, with the bulk volatility and oxidation state matching values of important known products. More complex systems such as diesel and wood-smoke vapors are characterized by progressively increasing mean carbon oxidation states, as expected based on thermodynamic constraints. However, the more complex systems in particular also show evidence for extensive fragmentation of the carbon backbones during SOA formation (Presto et al., 2009; Kroll et al., 2009).

The 2-D space allows us to constrain average organic properties much more tightly than our earlier 1-D-VBS (Donahue et al., 2006). The oxidation dimension gives us better predictions of the total organic mass (OM/OC) as well as OA properties, such as solubility, that are important to climate and perhaps health effects. We are thus developing a prognostic 2-D-VBS to efficiently model organic-aerosol chemical aging (Murphy et al., 2011). Full development of a 2-D-VBS model will require parameterization of oxidation kinetics well beyond the first one or two generations of oxidation chemistry, where organic behavior is relatively well understood (Atkinson, 1997; Jordan et al., 2008), to later generations where the chemistry is uncertain.

While the discussion in this work has focused on gas-phase chemistry, thermodynamic frameworks are agnostic about which phase chemical evolution occurs in. Thermodynamics are path independent. Thus, heterogeneous, condensed-phase, and aqueous-phase processes can be accommodated. Furthermore, as viscosity, molar weight, volatility, and molecular polarity are inter-related, we plan to treat condensed-phase dynamics in the future as well.

The implications of this work are summarized in Fig. 7, which shows hypothetical reaction trajectories for two model systems. This figure selects only two carbon isopleths but shows each oxygen isopleth for the mean group contribution method we will use for our prognostic 2-D-VBS; the oxygen isopleths represent a sort of generation count, at least before the onset of fragmentation in the reaction mechanism. Our challenge is to measure and model both the mean trajectories (shown with arrows) but more importantly the full...
breadth of these trajectories as well as the kinetics – the rate at which organics progress through the trajectories. Both trajectories show increasing fragmentation as $O_3$ rises, consistent with Kroll et al. (2011), and both trajectories (especially the $C_{10}$) will involve thousands of individual compounds as the mechanism pushes into later generations (Donahue et al., 2005; Aumont et al., 2005). Especially in the non-traditional trajectory, heterogeneous chemistry will clearly be of major importance, and in both cases accretion reactions, with the possibility of adding to the product carbon numbers, add yet another layer of complexity to the challenge. However, whatever mechanisms are ultimately responsible for delivering carbon to the upper left region of Fig. 7 in the atmosphere (forming LV-OOA), one conclusion that is unavoidable is that extensive oxidation chemistry must be involved.

Supplementary material related to this article is available online at:
http://www.atmos-chem-phys.net/12/615/2012/acp-12-615-2012-supplement.pdf.

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2-D VBS diagnostics

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