Fourier Transform Mass Spectrometry for the Molecular Level Characterization of Natural Organic Matter: Instrument Capabilities, Applications, and Limitations

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1. Introduction

All living matter in the environment (i.e., animals, plants, microorganisms, etc.) eventually dies and decomposes into what is known as natural organic matter (NOM). NOM is formed from a vast variety of sources that have been chemically or microbially degraded in the environment where they arise, and NOM can be generally described as a complex mixture of organic compounds (Stevenson, 1994). Within this mixture, some compounds retain their individual reactivity and characteristics, while others tend to aggregate together and act as a polymeric unit. Overall, NOM can encompass a variety of natural biomolecules, such as lipids, peptides/protein, amino-sugars, carbohydrates, lignins, tannins, and condensed aromatics. Because NOM is a random assortment of organic constituents, its size, shape, concentration, and other physico-chemical properties vary greatly with location and season. For these reasons, the molecular level characterization of NOM continues to be one of the greatest challenges to modern analytical chemists.

NOM is ubiquitously present in all natural waters, soils, sediments, and air, giving NOM a central role in numerous environmental processes. These processes are linked together by the global carbon cycle, which describes the storage and flux of carbon sources and sinks throughout the environment (Thurman, 1985; Eglinton and Repeta, 2003; Perdue and Ritchie, 2003). Special attention is generally paid to land-sea interfaces, atmosphere-sea interfaces, and long-term carbon burial/storage. NOM in soils affects the cation exchange capacity and water retention of soils, which has triggered studies by the agricultural communities. Furthermore, NOM in soils/sediments influences carbon sequestration and burial, and this carbon is altered over long periods of time and can be transformed to petroleum precursors. NOM in soils and rivers can affect the solubility, transport, and eventual fate of anthropogenic pollutants. These hydrophobic organic contaminants can interact and bind with NOM in the environment, making it difficult to trace throughout the river systems that eventually lead to the ocean. The amount of carbon in dissolved organic matter (DOM) in the ocean is approximately the same as that of atmospheric CO₂ (Hedges, 1992; Eglinton and Repeta, 2003) and this exchange has been directly linked to climate change (Canadell et al., 2007; Sabine and Feely, 2007). NOM in the atmosphere can exist as an aerosol or particulate, which impacts human health, climate, and overall air quality. The
brief explanation of NOM, along with the short list of NOM implications given here, barely touches upon the large variety of important research that is ongoing within the NOM community. The significance of NOM in the environment and the numerous roles that NOM plays in the biogeochemical processes that mediate Earth’s ecosystems highlight the necessity for a more fundamental comprehension of NOM chemistry and composition. The characterization of NOM from different sources is a difficult task, simply because NOM does not have an exact composition or structure and occurs at aqueous concentrations that can vary by 3 orders of magnitude, depending on NOM type and location. Bulk characteristics of NOM can be identified by elemental analysis, ultraviolet and infrared spectroscopy, and traditional one-dimensional nuclear magnetic resonance (Hatcher et al., 2001). Much has been learned about NOM chemistry from these techniques, such as elemental compositions and ratios (%C, %N, %O, %S, C/N, etc.), specific functional groups that primarily exist in NOM, and general trends that occur amongst different NOM samples. Chromatographic techniques, such as gas and liquid chromatography coupled to various detectors (i.e., flame ionization detection, mass spectrometry, photodiode array detection) has also provided a wealth of structural information on various types of NOM. Two-dimensional nuclear magnetic resonance has more recently been utilized to characterize and understand the nature of both soluble and insoluble NOM beyond that of the one-dimensional approach, and this has led to the ability to link NOM to well known biopolymer classes. However, despite the abundance of data that has been acquired using the aforementioned techniques, NOM remains as an analytical challenge. Because NOM exists amongst a background matrix, it can be difficult to separate from water or inorganic matter, without losing or altering the NOM (Mopper et al., 2007; Dittmar et al., 2008). Sample preparation for NOM is an important consideration. Furthermore, NOM is not amenable to most instrumental analyses, because it is a low concentration of highly functionalized polymeric substances that do not have uniform behaviour. NOM has a wide size and volatility range, since portions are hydrophilic, allowing them to be water soluble, while other parts retain their hydrophobic nature. Overall, the goal of molecular level characterization of NOM continues be a daunting task. The advent of atmospheric pressure ionization (API) sources and Fourier transform ion cyclotron resonance mass spectrometry (FTICR-MS) has revolutionized our ability to analyze NOM. These methods were originally employed by the biochemical communities to elucidate the structure of biological macromolecules (proteins, metabolic products, DNA, etc.). The knowledge that has evolved from these studies has been applied to NOM, where the goal is to transfer polar analytes in solution to molecular ions that can be detected by mass spectrometry. With the exception of petroleum and crude oil samples, the first application and use of electrospray ionization (ESI) for studying NOM was by McIntyre et al. (1997), utilizing a triple quadrupole mass spectrometer to analyze organic acids. Not long after, Fievre et al. (1997) utilized ESI-FTICR-MS to investigate the size and composition of humic and fulvic acids. In the last 15 years, the utilization of FTICR-MS for the analysis of NOM samples has increased from year to year, and from 2000 to 2010, the number of publications has increased by nearly 900% (Fig. 1). Clearly, FTICR-MS is a powerful technique that has great promise in the field of NOM chemistry, with a rise in its use progressing every year. In this chapter, the spotlight lies in utilizing FTICR-MS specifically for the characterization of NOM, concentrating on particular instrumental capabilities, its application to a variety of different types of NOM, and limitations that exist for the acquisition and analysis of data.
Fig. 1. Approximate number of publications on the topic of natural organic matter (water, soil, sediment, aerosol, petroleum, crude oil, etc.) analysis using Fourier transform ion cyclotron resonance mass spectrometry. Search conducted in September, 2010

2. Introduction to Fourier transform ion cyclotron resonance mass spectrometry

Over the course of the last decade, FTICR-MS has emerged as an invaluable tool for the characterization of NOM by providing details about its composition. While previous chemical and instrumental analyses (e.g., gas and liquid chromatography, ultraviolet and infrared spectroscopy, fluorescence excitation emission matrix spectroscopy, nuclear magnetic resonance (NMR), elemental and isotopic analyses, etc.) have revealed vital information about NOM, these techniques are biased for certain compound classes and fail to resolve the numerous constituents in NOM (Hatcher et al., 2001; Leenheer and Croué, 2003). Other mass spectrometers have been employed for the analysis of NOM, such as quadrupoles, ion traps, and quadrupole time of flights. However, FTICR-MS has abilities and advantages over these systems and has achieved a more in depth analysis of NOM where other mass spectrometers have been marginally or less successful (Sleighter and Hatcher, 2007).

The detailed theory and instrumental parameters of FTICR-MS are expertly reviewed by Marshall et al. (1998). Here, we give a brief overview of the instrument. Ions are produced in the ion source region (via a variety of available ion sources discussed in section 3) that is maintained at atmospheric pressure. These ions are focused, using ion funnels and skimmers, into differential pumping regions that vary in pressure from atmospheric pressure, to low vacuum ($10^{-4} - 10^{-6}$ mbar) just after the ion source region, to high vacuum.
(10^{-9} - 10^{-10} \text{ mbar}) in the ICR cell. Ions are steered through these regions and typically pass through a mass analyzer in the lower vacuum area where an initial sort and storage occurs. Commercial instruments vary in which type of mass analyzer is used; some are a combination of hexapoles and/or quadrupoles, while others are linear ion traps. In this first mass analyzer region, only ions of a certain mass to charge (m/z) range (typically 100-2000 m/z) are allowed to be accumulated for a designated time prior to their transfer to the detector. Once past this accumulation stage, the ions are guided through more pumping stages of the ion transfer optics region and are eventually transferred into the horizontal bore of a cryogenic magnet, where they are trapped in the ion cyclotron resonance cell.

Modern commercial FTICR-MS instruments employ cryomagnets of various strengths (usually 7-15 T). Once ions are trapped in the ICR cell under the influence of a homogeneous magnetic field, they circulate at a frequency characteristic of their m/z value. Cyclotron frequency is inversely proportional to m/z, as shown below:

\[ f_c = B_0 \frac{z}{m} \]  

(1)

In equation 1, \( f_c \) is the cyclotron frequency, \( B_0 \) is the magnetic field strength, \( z \) is the charge of the ion, and \( m \) is the mass of the ion. Ions in orbit in the ICR cell are excited by a broadband RF pulse, which increases the radius of their orbit but not their frequency. The ions are now circulating closer to the detector plates, where the ion packets can induce an image current on the receiving electrode. Field inhomogeneities cause the ions to lose coherence and orbit radius, which leads to collapse of the ions to the central core of the ICR cell. This produces an image current trace that is called a free induction decay (FID), similar to what is observed for NMR signals. The time-domain FID signal (Fig. 2a) is digitized and Fourier transformed into a frequency domain signal (Fig. 2b), after acquiring and summing multiple FID spectra to build up signal-to-noise. The frequency domain data are then converted into mass spectra (Fig. 2c) by use of equation 1 and calibrated with compound mixtures having components with known m/z values. This sequence of data detection and conversion is shown in Fig. 2 for a sample of Suwannee River NOM.

In our opinion, FTICR-MS is the only type of mass spectrometer that can achieve the resolving powers that are necessary (those exceeding 10^5) for mass deconvolution of the thousands of compounds that are present in a single NOM sample (Fig. 2c). Orbitrap mass spectrometers (providing resolving powers of approximately 60,000) provide sufficient resolution for evaluating elemental formulas of mixtures containing only C, H, and O molecules (see section 5 and Fig. 8), but the inclusion of heteroatoms in the molecules (i.e., N, S, and P) requires the use of FTICR-MS to achieve assignment of exact elemental formulas. Fig. 2d shows that the most intense peaks are detected at odd nominal masses, indicating that they are composed of compounds with either 0 or an even number of nitrogens (based on the nitrogen rule). Ions detected at even nominal masses contain an odd number of nitrogens or are \(^{13}\text{C}\) isotopologues of the \(^{12}\text{C}\) compounds detected (Fig. 2e). Based on the identification of the \(^{12}\text{C}\) and \(^{13}\text{C}\) isotopologues, one can determine the charge state of the compound. If a compound is singly charged, then the \(^{13}\text{C}\) isotope peak will be observed at 1.0034 mass units higher than the \(^{12}\text{C}\) peak, which is the mass difference between \(^{12}\text{C}\) and \(^{13}\text{C}\). Doubly charged peaks have isotopes that appear at 0.5017 mass units higher, but doubly charged peaks are rarely detected in NOM samples (Kujawinski et al. 2002; Stenson et al., 2002; Kim et al., 2003). Generally, at least 10 peaks are detected at each individual nominal mass, with upwards of 20-30 being observed in some cases.
Fig. 2. Suwannee River NOM analyzed by negative ion mode ESI-FTICR-MS: a) time domain free induction decay (FID), b) FID Fourier transformed into a frequency domain spectrum, c) mass spectrum after frequency is converted to m/z, d) expanded region of mass spectrum at 371-398 m/z, e) nominal masses 375 and 376 of the mass spectrum. Numbered peaks in (e) correspond to Table 1.

The ultrahigh resolving powers that FTICR-MS at 12 T routinely achieves is highlighted in Fig. 2e and is the main reason why FTICR-MS is preferred over other mass spectrometers that exhibit only nominal mass resolution (see examples: Fig. 3 in Kujawinski et al., 2002;
The equation for calculating resolving power is shown below in equation 2, where RP is resolving power, m is mass, and FWHM is the full width at half maximum of the peak.

\[ P = \frac{m}{\text{FWHM}} \quad (2) \]

**Table 1. Details of the peaks shown in Fig. 1e**

Resolving power calculations are demonstrated in Table 1, which shows the molecular formula assignments to the peaks in Fig. 2e, along with measured details of each. Because the frequencies, at which ions orbit within the ICR cell, can be measured very accurately, m/z can also be calculated very accurately, usually to the fifth decimal place. With careful external and internal calibration (Sleight et al., 2008), accurate m/z values can be calculated and utilized for the determination of unique molecular formulas. These can be confidently assigned with an error difference (between the measured m/z and the calculated exact m/z) of less than 0.5 ppm (or 500 ppb). Once molecular formulas are...
assigned to the majority of the peaks in the mass spectrum, compositional make-up of the sample can be established.

3. FTICR-MS instrument capabilities and data acquisition

While FTICR-MS is a particularly impressive instrument and has the ability to provide molecular level details about NOM samples, there are many factors to take into account during data acquisition. It is important to consider the specific instrument’s design and capabilities, in order to optimize certain parameters and decide how the data should be acquired so that high quality, meaningful mass spectra are obtained.

The first concern is for sample composition. There are several inherent difficulties with obtaining publishable mass spectra of NOM, particularly when the organic matter exists at a low concentration in the presence of a much higher concentration of inorganic matrix components. Salty samples are especially problematic, as emphasized in Fig. 3, which shows a NOM sample from the Elizabeth River that was analyzed by ESI-FTICR-MS before and after desalting by electrodialysis. An expanded region of the mass spectra acquired for the Elizabeth River NOM before desalting (Fig. 3a) shows high magnitude peaks with a high mass defect (0.7-0.8) and lower magnitude peaks at mass defects that are typical of organic matter (0.0-0.5) in this size range. Once the riverine NOM has been desalted (Fig. 3b), the high mass defect peaks attributed to salts are absent and the lower mass defect peaks are enhanced. Mass defect refers to the deviation of an m/z value from the exact nominal mass, and it is indicative of the type of compound present, based on the mass defect of atoms in organic compounds. The exact masses (in amu) of $^{12}$C, $^1$H, and $^{16}$O are 12.000000, 1.007825, and 15.994915, respectively. Hydrogen has a positive mass defect, while oxygen has a negative mass defect. Thus compounds that are oxygen-rich and/or hydrogen-poor will display peaks at a lower mass defect (ca. 0.0-0.2), while compounds that are hydrogen-rich and/or oxygen-poor give peaks with a higher mass defect (ca. 0.2-0.5). The composition of peaks detected at high mass defect in Fig. 3a has not been confidently determined, but they are inorganic compounds present amongst the NOM sample. Their magnitudes are intense because they have higher ionization efficiencies than the OM constituents. Compounds that exist as ions in solution will ionize much more readily by ESI than compounds that do not. ESI is a competitive ionization process, and the OM components simply cannot out-compete the inorganic compounds for the negative charge, hence explaining why OM peaks are detected at lower magnitudes than the inorganic compounds. Because the OM compounds are the analytes of interest, desalting techniques are utilized to remove the background of inorganic matrix.

There is constant research being performed to determine the best method for desalting, isolating, and concentrating NOM samples. Traditionally, NOM was extracted from water, soils, and sediments as humic substances, which can be further categorized as fulvic acid, humic acid, and humin, depending on the pH at which they are soluble (Stevenson, 1994). Humic substances are extracted by well established acid/base laboratory protocols (Schnitzer and Khan, 1978; Thurman and Malcolm, 1981). More recently, DOM has been the focus of many studies interested in carbon cycling through groundwater, porewater, rivers, estuaries, and the ocean. Methods that are commonly employed to isolate the NOM from water are ultrafiltration, solid phase extraction, electrodialysis, and combined reverse osmosis electrodialysis. Each of these methods has certain problems associated, such as irreversible NOM sorption (lowering the NOM recovery of that technique), breakthrough
Fig. 3. Elizabeth River NOM analyzed by negative ion mode ESI-FTICR-MS before (a) and after (b) complete desalting by electrodialysis (expanded ranges are 450-500 m/z and 466.7-467.4 m/z). Peaks with high mass defects (0.7-0.9) are from incomplete desalting. More NOM peaks are detected when these compounds are not competing for a charge.
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contamination/bleeding, resin/membrane contamination of the NOM, and typically time consuming cleaning requirements (Simjouw et al., 2005; Mopper et al., 2007; Dittmar et al., 2008). These methods are biased by the chemical or physical properties that regulate the extraction procedure, and readers are referred to more in depth discussions of these desalting techniques, along with the pros and cons of each, in the references given above. In general, desalting and sample cleanup are important issues to consider when preparing samples for FTICR-MS.

After the appropriate sample preparation is performed, the optimal ionization source should be identified. There are many different commercially available API sources for the application to NOM (Hoffmann and Stroobant, 2003), such as ESI (Bruins, 1991; Gaskell et al., 1997; Cech and Enke, 2001), chemical ionization (CI; Bruins, 1991; Harrison, 1992), and atmospheric pressure photoionization (APPI; Raffaelli and Saba, 2003; Bos et al., 2006; Purcell et al., 2007). Each method varies in its ionization mechanism, and, consequently, the analytical window for each is quite different. It is important to mention that non-ionizable compounds that exist in the NOM will be invisible to the mass spectrometer. This is important, because each ion source has its own innate bias and each can give a different resulting mass spectrum for the same NOM sample (Hockaday et al., 2009). The ionization methods mentioned above are known as ‘soft’, meaning that compounds are not fragmented (as they are in electron ionization) and molecular ions (M+\(^+\), M-\(^-\)) or pseudomolecular ions, also known as molecular ion adducts [(M+H)+, (M+Na)+, (M-H)-, (M-Cl)-], are predominantly observed in the mass spectrum. Mechanistic studies of each of the ion sources are referenced above, but brief explanations of each are described here.

CI introduces a reagent gas (methane is quite common, CH\(_4\)) into the ion source to produce primary ions of the reagent gas (i.e., CH\(_4\)+) to collide with the molecule of interest. Through ion-molecule collisions, where the reagent gas ion acts as a Brønsted-Lowry acid and the analyte is a Brønsted-Lowry base, and proton transfer reactions, the analyte is ionized with minimal fragmentation. During ESI, the liquid sample is sprayed through a needle, and a high voltage difference between the spray needle and metal inlet induces a charge on the sprayed droplets. The charged droplet diminishes in size as the solvent is evaporated (by aid of heat or a drying gas), concentrating the charges held on the droplet. As charge-charge repulsions occur, the Rayleigh limit is exceeded, making the Coulombic repulsions greater than the surface tension of the droplet. The result is that the droplet bursts into many smaller droplets that can be completely desolvated, leaving only charged analyte ions in the gas phase for further introduction into the mass spectrometer. ESI operates in either positive or negative mode, depending on the functional group composition of the analyte. Functional groups that will readily lose a proton (such as alcohols, carboxylic acids, cyanides, peptides, nitric- and sulfonic- acids, and phosphates) are analyzed in negative ion mode. Basic functional groups that can easily gain a proton (i.e., amines, amides, peptides, and thiols) are analyzed in positive ion mode. By changing the pH of the sample solution (slightly basic for negative ion and slightly positive for positive ion), one can increase the ionization efficiencies for ESI. Another ion source of choice for analyzing less polar compounds is APPI, and APPI does not tend to suffer from charge competition with inorganic matrices, which is commonly observed in ESI. In APPI, ionization is initiated by supplying UV photons to the analyte molecule (typically via a krypton lamp). The analyte absorbs the photons and enters into an excited state. The analyte becomes ionized when the energy of the UV photons is greater than the ionization energy of the analyte. Dopants (such as toluene or tetrahydrofuran) are usually employed to act as intermediates between the
photons and analytes, so that charge exchange and proton transfer reactions can occur more readily, increasing the ionization of NOM molecules.

Overall, before selecting an ion source, it is beneficial to know the bulk functional group composition of the sample (by previously obtained FTIR or NMR data), because then an informed decision can be made on which ion source(s) to employ for that specific sample. ESI and APPI are the most commonly used ion sources for the analysis of NOM, and Hockaday et al. (2009) performed a study to investigate which appeared to be optimal for a terrestrial sample obtained in the Dismal Swamp (Suffolk, VA, USA). They found that little overlap existed between the formulas assigned to APPI(+) and ESI(+ and -) mass spectral peaks, suggesting that data acquired from the two ion sources complemented each other greatly. Furthermore, APPI yielded formulas that were more aromatic and less polar than those from ESI, which is expected due to the ionization mechanism for each. Based on Hockaday et al. (2009) and the discussion of the types of compounds ionized by various ionization sources, many investigations of NOM do multiple-source FTICR-MS analyses to produce data that supplement each other, furthering the overall characterization of the NOM sample.

Once the best ion source has been determined for a particular sample, the mass spectrometric parameters can be examined. The two main ways to acquire mass spectral data are either broadband mode or narrow scan mode. The vast majority of publications that analyze NOM by FTICR-MS utilize broadband mode, where all m/z values, across a wide range of generally 200-2000 m/z, are detected during the analysis. Narrow scan mode is commonly referred to as sequential selective ion accumulation (SSIA; Sleighter et al., 2009) or as selected ion monitoring (SIM; Kido Soule et al., 2010). During SSIA (or SIM), the initial mass analyzer [quadrupole for Sleighter et al. (2009) and linear ion trap for Kido Soule et al. (2010)] isolates ions within a narrow range of m/z values before transferring the ion packet to the ICR cell. The operator selects the range, and ions outside of this range will be eliminated by the initial mass analyzer, thus decreasing the total number of ions in the ICR cell simultaneously. Because there are fewer ions in the cell at the same time, space-charge effects are minimized, generally increasing the resolving power and selectively enhancing the S/N of the peaks. Using SSIA, the sample is analyzed multiple times, incrementally increasing the m/z range, so that the entire mass range is eventually covered. By this SSIA method, the m/z ranges are acquired in sequential ‘slices’, and these slices can be merged together to assemble the entire mass spectrum.

Whole water from the Dismal Swamp (Suffolk, VA, USA) was sterile filtered (0.2 μm) to remove particulates and bacteria and analyzed directly using both broadband and SSIA modes, as shown in Fig. 4. The colors in Fig. 4b show each ‘slice’ that was acquired, and the ‘slices’ overlap by approximately 30-40 m/z to ensure that no area is missed. The nominal mass region in Fig. 4 is shown in order to highlight the increase in S/N and resolving power. An S/N threshold of 3 is commonly used for peak-picking, and the increase in overall S/N using SSIA leads to a larger number of peaks detected. Table 2 gives details for the mass spectra shown in Fig. 4, parsed according to the m/z range that was acquired. This table shows that the increase in the number of peaks, S/N, and generally resolving power exists for SSIA across the entire mass range and is more substantial and pronounced at higher m/z values. The broadband analysis requires 20 minutes (1.0 sec ion accumulation and 200 co-added scans), while the SSIA analysis requires 30 minutes (1.0 sec ion accumulation and 50 co-added scans, for 6 separate m/z ranges). While the SSIA mode involves more time to acquire data, 2880 more peaks were detected during that time. It is important to note that while nearly 3000 more peaks were detected with SSIA, this does not
necessarily translate to 3000 more molecular formulas being assigned, because some of the extra peaks detected by SSIA are isotopic peaks.

Fig. 4. Dismal Swamp whole water analyzed by negative ion mode ESI-FTICR-MS in broadband mode (a) and using sequential selective ion accumulation (b). The insets show expanded mass spectra at m/z 540-700 and 425.0-425.35, to show the peaks at higher m/z more clearly and to highlight the enhanced S/N achieved using SSIA.
Table 2. The number of peaks detected, average S/N, and average resolving power for each m/z range of the broadband mass spectra and SSIA mass spectra shown in Fig. 4.

The discussion in this section of sample preparation, ion source selection, and optimizing FTICR-MS acquisition modes emphasizes just a few factors that are of paramount importance when analyzing NOM. There are countless other mass spectral parameters that can also be optimized (ion source voltages, ion optics and transmission parameters, ion accumulation times, trapping parameters within the ICR cell, etc.) in order to obtain the highest quality data. However, these parameters are instrument specific, and each individual instrument requires tuning before each sample set is analyzed. Further discussion and advice on these parameters is generally reviewed in detail by the manufacturer and is beyond the scope of this chapter.

### 4. FTICR-MS applications to NOM

As discussed above, molecular formulas can be assigned to the multitude of peaks detected in mass spectra of NOM, and this capability is the main justification for use of FTICR-MS for the molecular characterization of various NOM samples. The molecular formulas provide meaningful compositional information that is associated with groups of natural biopolymers. This practice of correlating assigned formulas to sample composition has been applied to various types of NOM, and several methods exist to assist in this correlation. Because it can be difficult, tedious, and labor-intensive to compare the thousands of assigned molecular formulas for a single NOM sample, generally visualization diagrams are called upon to assist in displaying the formulas in a chemically representative manner. The two-dimensional van Krevelen diagram has been the most commonly used approach, which plots H/C values vs. O/C values (van Krevelen, 1950; Kim et al., 2003). Each molecular formula aligns on the diagram in a location that can typically be correlated to that commonly associated with natural biomolecules, as shown in Fig. 5 for riverine DOM isolated from the Elizabeth River in south-eastern Virginia, USA. The circles overlain on the plot highlight the types of molecules that are commonly detected in NOM samples, as well as their position on the van Krevelen diagram, based on the compound’s elemental ratios (Kim et al., 2003; Sleighter and Hatcher, 2007; Hockaday et al., 2009). It should be noted that these circles are not strictly representative of all similar molecules, but rather approximate...
guidelines for identifying compounds of similar composition. Relating formulas to compound classes in this manner has been exploited in many studies of various types of NOM over the years, and interested readers are referred to the literature for more details (Sleighter and Hatcher, 2007; Reemtsma, 2009; references therein).

Another diagram that is often used for NOM characterization is the Kendrick mass defect (KMD) plot (Kendrick, 1963). KMD analysis converts m/z values to Kendrick mass values by multiplying the m/z value by the ratio of the nominal mass CH\(_2\) group (14.00000) to the exact mass of a CH\(_2\) group (14.01565), as shown below in equation 3. Then, KMD is determined by subtracting the nominal Kendrick mass (KM) from KM, as shown in equation 4. KMD values can then be plotted against their nominal KM values (as shown in Fig. 6), and formulas with the same KMD, those falling on a horizontal line, differ only by a CH\(_2\) group (or multiple CH\(_2\) groups). KMD values increase with the number of added H atoms, thus aliphatic compounds will have high KMD values and aromatic compounds will have lower KMD values. This coincides with mass spectral data expanded at individual nominal masses, where m/z values with low mass defects are hydrogen-poor and m/z values with higher mass defects are hydrogen-rich (as shown in Fig. 2e and Table 1).

\[
\text{Kendrick Mass (KM)} = \frac{m/z \text{ value}}{14.00000} \times 14.01565
\]

(3)
Kendrick Mass Defect (KMD) = KM - nominal KM

Originally, KMD was utilized for assigning molecular formulas, by establishing homologous CH$_2$ series that could be expanded from low m/z to high m/z (Stenson et al., 2003). Peaks at low m/z can more easily be assigned a molecular formula because fewer formulas exist within the selected error limit (usually 0.5 ppm). Generally, only 1 molecular formula exists within this error for peaks less than 500 m/z, but beyond this value, multiple formulas are possible. Once molecular formulas are unambiguously assigned to peaks of low m/z, peaks at high m/z values that have multiple formula choices can be related to formulas assigned at lower mass by assuming that they belong to a CH$_2$ homologous series. If one of the formulas belongs to a homologous series, then it is very likely the correct formula, and the others can be eliminated. This approach to formula assignment is called ‘formula extension’ and can be performed manually or written into software designed to assist in formula assignment (Kujawinski and Behn, 2006; Grinhut et al., 2010). While CH$_2$ is the most commonly used group, other functional groups can be utilized (i.e., OCH$_2$, COO, O, H$_2$O, H$_2$, etc.) depending upon the make-up of the sample (Sleighter and Hatcher, 2007).

Because of the recent growth in use of FTICR-MS for the analysis of NOM, articles have been published summarizing the findings of these studies and making suggestions for future work (Sleighter and Hatcher, 2007; Reemtsma, 2009). Most recently, multivariate statistical analysis in combination with visualization diagrams have been utilized to evaluate relationships among sample sets. Hierarchal cluster analysis (HCA) seeks correlations among samples displayed in a data matrix and illustrates the results in a hierarchical tree, or a dendrogram, where the branching reveals the similarity among the samples. HCA has been utilized in numerous studies, one of which shows that there are no significant differences between the formulas assigned to a depth profile of DOM from the Weddell Sea (Koch et al., 2005). In other studies, Dittmar et al. (2007) compared DOM from coastal mangrove forests (before and after photo-irradiation) to open ocean seawater. They found that photo-degraded mangrove DOM becomes similar in composition to open ocean seawater DOM. Koch et al. (2008) evaluated the various fractions of DOM collected from reversed phase HPLC separations, and Schmidt et al. (2009) used HCA statistical correlations to differentiate between pore water DOM and riverine DOM. While HCA is very useful for grouping samples based on their similarity, it does not indicate the reasons why the samples are similar or different. The variation between the samples is not explained, and another method is required for this determination.

By combining HCA with another statistical method, the variance between samples can be elucidated. Kujawinski et al. (2009) employed HCA combined with both non-metric multidimensional scaling (NMS) and indicator species analysis (ISA) to optimize the a priori grouping of samples for subsequent ISA, where specific mass spectral m/z values can be identified as an indicator species for a certain group of samples. Once all the indicator species for the samples are identified, then molecular formulas were examined more closely. Indicator species for surface ocean DOM samples were speculated to be biologically-derived and to represent a more labile component of the marine DOM pool, while indicator species for riverine/estuarine DOM were found to be similar in composition to lignin-derived species that have been linked to terrestrial-sourced DOM. Bhatia et al. (2010) also used these multivariate statistical methods recently to characterize DOM from the Greenland ice sheet and were able to link subglacial, supraglacial, and proglacial DOM to various allochthonous and autochthonous sources and processes. Another recent study utilizing ISA...
distinguished compounds that are specific to various fractions of NOM in soils and crop biomass (Ohno et al., 2010). It was discovered that water extractable OM from plant biomass had marker components that could be classified as lipids, proteins, carbohydrates, lignin, and unsaturated hydrocarbons, while the water extractable OM from soils contained more lignin- and carbohydrate-sourced compounds. The mobile humic acid extract of soils displayed mostly lignin-like markers and the immobile humic acid markers clustered in the condensed aromatic space. A general trend of increasing aromaticity (i.e., decreasing H/C ratio) was observed along the humification gradient, from plant biomass to water-extractable soil OM to refractory, stabilized humic acids.

Fig. 6. Kendrick mass defect plot (using a CH$_2$ group) for Elizabeth River NOM isolated by small-scale electrodialysis and subsequent analysis by negative ion mode ESI-FTICR-MS. The yellow box is expanded in the lower diagram to highlight the points that fall on a horizontal line, indicating that they are part of a homologous CH$_2$ series.
By combining HCA with PCA, Hur et al. (2010) developed a method for comparing multiple petroleum samples that were analyzed by APPI-FTICR-MS. Using these multivariate statistical tools, twenty petroleum samples could be compartmentalized into numerous compositional groups, such as those enriched in hydrocarbons, oxygen series (O1, O2), nitrogen series (N1, N2O1), or sulfur series (S1, S2, O2S1). Sleighter et al. (2010) also employed a combined approach of HCA and PCA, but in this study the goal was to characterize 38 NOM samples along a terrestrial to marine transect of the lower Chesapeake Bay and coastal Atlantic Ocean, using samples that were prepared for mass spectral analysis by either sterile filtration only, solid phase C18 extraction, or small-scale electrodialysis. Not only were differences detected between NOM samples from various locations, but it was also found that the method of preparation for the same NOM sample changed its composition, as determined by ESI-FTICR-MS. Terrestrial samples contained lignin, tannin, and condensed aromatic structures in high relative magnitude, while marine DOM was composed more of aliphatic and lignin-like compounds, as well as compounds containing more heteroatom (NSP) functionalities. Samples desalted by electrodialysis retained the more polar compounds (tannins, carbohydrates, and those containing heteroatoms) that were eliminated during the C18 extraction procedure. Overall, these recent developments in NOM characterization, made by exploiting multivariate statistical analyses, highlight the direction in which the NOM community is progressing. It is very likely that statistical analyses will become more commonplace, especially as researchers continue to accumulate large sample sets that would be otherwise very difficult to analyze manually.

5. Limitations of FTICR-MS

A major concern regarding the use of FTICR-MS for NOM characterization in recent years has been establishing an appropriate instrumental/solvent blank, to evaluate the background peaks detected in the mass spectra. Most analytical techniques have a straightforward manner for nulling the blanks, such as a basic background subtraction of the blank analysis from the sample analysis. Unfortunately, this method cannot be applied so simply when using a competitive ionization source such as ESI. Most NOM samples are analyzed in a mixture of methanol and water, thus a clean solvent blank of methanol and water is typically evaluated before analyzing samples. However, any contaminants present within the instrument itself will not have any analyte molecules with which to compete for charge. Thus, these substances acquire most of the charge, giving them an enhanced magnitude in the mass spectra. Once a sample is introduced, the analyte molecules out-compete the contaminants for the charge in most cases, and then those contaminants are only observed at lower absolute and relative magnitudes, as shown for Mount Rainier humic acid in Fig. 7. The NOM sample is analyzed at various concentrations, to determine which concentration is sufficient to out-compete most (if not all) of the contaminants present in the FTICR-MS. For peaks that are detected in both the blank and the NOM sample, it is difficult to determine if that peak is actually in the sample or if it is due to contamination. In the case of Fig. 7, m/z 415.322 is detected at nearly a constant magnitude in the instrument/solvent blank and at each concentration of Mount Rainier humic acid. Because the other contaminants (at 414.8–414.9 and 415.27) are nearly eliminated at 50 mg/L OM, it is likely that m/z 415.322 is present in the NOM sample, since it continues to be detected at all NOM concentrations. Nonetheless, some researchers err on the side of caution and argue that this peak should be removed from further analysis, since it is detected in the blank as well. Only 1 nominal mass region is highlighted in Fig. 7 as an example, but this trend in blank
Dilution/elimination occurs across the entire mass spectral range of 200-800 m/z. Thus, it is common to question the existence of several hundred peaks that are detected in each NOM sample that also overlap with the peaks in the blank. There is not currently a standard protocol among researchers for determining when to remove peaks from analysis, but typically the most conservative approach is taken and overlapping peaks in the blank and sample are removed from consideration as analyte peaks. It is important, though, to understand the overlap between solvent blank and sample analysis, so that peaks that are important to the NOM composition are not considered artifacts of the instrument itself.

Another concern, when analyzing samples as complex as NOM, is peak reproducibility. The vast majority of published studies only analyze each sample once, rather than in replicate. This is due to sample throughput and cost. However, more recently, there has been more concern for reproducibility in order to characterize NOM reliably. The sample must be analyzed, in either duplicate or triplicate, to ensure reproducibility. This facilitates comparison of solvent or instrument blanks to replicate injections, enabling decisions as to which peaks should be included for further data analysis. Kidou Soule et al. (2009) discovered that broadband acquisitions offer the highest repeatability for peaks detected and peak height (along with highest throughput), when compared to SSIA (or SIM) acquisition. In a recent study, Hur et al. (2010) analyzed 20 petroleum samples by APPI in triplicate over the course of 2 consecutive days. They reported that the three mass spectra for each sample were quite consistent, showing standard deviations of less than 5% of the initial values. Sleighter et al. (2010) performed a study comparing a single sample analyzed on different days by replicate injections of Dismal Swamp (Suffolk, VA, USA) whole water over the course of 31 days. The mass spectra produced were visually very similar, and multivariate statistics was utilized to evaluate the reproducibility. Based on HCA and PCA, the different analyses were found to be virtually identical when compared to a variety of other samples.

These three studies highlight that ESI-FTICR-MS has the potential to be very reproducible and reliable, when the same instrumental parameters are utilized. However, no studies have tested the repeatability for various instrument operators or for the same sample analyzed on different FTICR-MS instruments.

Another unknown factor in the analysis is the fraction of the NOM that is identified by FTICR-MS. The ion sources described here for coupling to FTICR-MS are well known to exhibit biases for certain types of molecules, depending on the ionization mechanism and the ionization efficiency of the analyte in the midst of a complex matrix. If only 5% of the NOM is ‘observed’ by FTICR-MS, rather than 50% (or perhaps even higher), then the implications are enormous for complete characterization. Based on this concern, Hockaday et al. (2009) suggest that an internal standard be made widely available for spiking into NOM samples prior to FTICR-MS analysis. Using this approach, the ion source performance on various instruments could be assessed and compared. Furthermore, this standard could also used for internal calibration, making datasets acquired on different instruments by various operators more comparable.

Once a high quality mass spectrum is obtained for a NOM sample, data analysis can proceed in order to obtain molecular level characterization. This characterization is performed by assigning the m/z values of peaks detected to molecular formulas, by using a molecular formula calculator. First, one must determine which atoms to consider for assignment. Typically C, H, O, N, and S are used, although some studies also include P, Na, and/or Cl. It is well known that the number of possible molecular formulas increases with increasing 1) permissible error difference between measured m/z and exact calculated m/z for the formula in question; 2) m/z value (higher m/z peaks are inherently less precisely measured); and 3) number of atoms used for formula assignment (Kim et al., 2006;
Kujawinski and Behn, 2006; Koch et al., 2007; Reemstma, 2009), as shown in Fig. 8. As one can see, increasing the allowable error and including more elements significantly increases the number of chemically possible formulas. The relationship between resolving power and the error difference between two formulas with slight differences in m/z values is reviewed in the four references given above, but Kim et al. (2006) reported that all theoretically possible elemental compositions of C,H,N,O,S up to 500 Da could be resolved at an accuracy of approximately 0.1 mDa (corresponding to a resolving power of 5,000,000 at m/z 500), allowing for a unique molecular formula to be assigned with confidence. While resolving powers of this magnitude are not currently routinely achieved during the analysis of NOM, it is likely that as technology continues to improve and FTICR-MS instruments at high magnetic fields are utilized, that these values will be achieved in the near future.

Fig. 7. Negative ion mode ESI-FTICR mass spectra of a) instrument/solvent blank (1:1 water:methanol with 0.1% ammonium hydroxide) and Mount Rainier humic acid dissolved at 5 (b), 25 (c), and 50 (d) mg/L organic matter in 1:1 water:methanol with 0.1% ammonium hydroxide. The broadband mass spectrum has been expanded at 414.7 – 415.5 m/z.
Fig. 8. The number of chemically possible molecular formulas for hypothetical m/z 499.21257 at various error values for the different elemental compositions specified in the legend.

The inherent complexity incurred while assigning a unique molecular formula to an individual m/z value is highlighted in Fig. 8, and this difficulty is amplified when the task at hand is to assign thousands of m/z values to formulas for a single NOM sample. This process can be very labor-intensive. However, by carefully calibrating the mass spectral data, lower error differences can be tolerated (0.5 ppm has become quite common in the literature), to minimize the number of formulas that match the measured mass within the selected error. There are also other methods to assist in the determination of molecular formulas, such as using KMD analysis and the ‘formula extension’ approach, as described above. Furthermore, establishing rules that the formulas must obey (as in Stubbins et al., 2010), to be assigned to a chemically relevant molecule, also helps to reduce the number of possible formulas. Formulas that are unlikely to be acceptable can be identified by examining the isotopic peak(s) as described by Koch et al. (2007) and featured in Fig. 9. Based on the natural abundance of $^{13}$C (1.1%), the number of carbons in the correct formula can be calculated from the relative magnitude of the $^{13}$C peak, eliminating incorrect formulas (Fig. 9a). In the case of Fig. 9b and 9c, C$_{20}$H$_{27}$O$_7^-$ is determined to be the correct formula for the peak, because the predicted relative abundance of its $^{13}$C isotopologue is much closer to what is detected in Fig. 9a than it is for the formula C$_{16}$H$_{30}$N$_3$O$_7$P$_1^-$ in Fig. 9c. For Fig. 9d, the peak at 487.30445 is not initially assigned a molecular formula, because Cl is not included in the formula assignment parameters. Upon closer inspection of the peak, the distinctive isotope pattern for Cl$^{-}$ adducts is observed and matches very closely to that of a simulated theoretical pattern for C$_{23}$H$_{38}$O$_6$Cl$_1^-$(Fig. 9e). Koch et al. (2007) cautions that the S/N of the isotopic peak should be at least 25, otherwise the deviation (between the number of carbons in the proposed formula and that calculated from the relative magnitude) is too great to be a reliable tool for formula elimination. In general, by ensuring that an accurate
Fig. 9. Coastal marine DOM isolated by C18 solid phase extraction and subsequent analysis by negative ion mode ESI-FTICR-MS, expanded at 379.0-380.2 m/z (a) and 487.0-490.5 m/z (d), along with the simulated isotope patterns for the possible formulas specified (b, c, e). Error values are in ppm.
calibration is achieved, setting a low error difference, utilizing KMD analysis and elemental ratio rules, and exploiting isotopic relative abundances, one can assign molecular formulas more easily and with additional confidence. Ideally, a generally accepted protocol should be established and each of these tools could be written into a standard software program, similar to that developed by Kujawinski and Behn (2006). Then data analysis consistency between samples would be more reliable, and human error introduced by different analysts could be minimized. This specific step has yet to be taken, but with emerging literature in which suggestions are made regarding formula assignment rules, some standardization will become reality in the near future.

Even though data analysis of FTICR mass spectra of NOM can provide accurate molecular formulas for the peaks detected, structural information for these peaks, which is often the key to understanding the source of the NOM, still remains elusive. Previous FTICR-MS studies focused on molecular characterization of DOM from terrestrial and marine waters have found that a significant overlap of the mass spectra is observed, as approximately 30% of the assigned formulas are shared between the two different types of DOM (Koch et al., 2005; Sleighter and Hatcher, 2008). However, mass spectrometry cannot distinguish structural isomers of an elemental composition, so the possibility exists that identifying the same elemental formula in multiple samples does not necessarily mean that their molecular structures also correspond. For these reasons, tandem mass spectrometry (MS/MS) has been employed to isolate and fragment ions in the mass spectrometer, and this can provide structural information. Because the ions detected during the mass spectral analysis of NOM are predominantly singly charged, only neutral losses (i.e., H$_2$O, COO, CO, and OCH$_2$) have been observed by MS/MS using FTICR-MS (Stenson et al., 2003; Reemtsma et al., 2008; Witt et al., 2009; Liu et al., 2011) and other lower resolution mass spectrometers (Fievre et al., 1997; Plancque et al., 2001; Leenheer et al., 2001; McIntyre et al., 2002). Furthermore, the complexity of NOM samples and the proximity of peaks detected at each nominal mass makes it particularly difficult to isolate a single peak in the mass spectrometer for fragmentation and subsequent detection of those fragments. While it is possible to do this, as shown by Witt et al. (2009), fragmenting each peak detected at every nominal mass from approximately 200-800 m/z by FTICR-MS/MS is quite a daunting task, one that will be labor intensive and require long instrument run times. The mass of data that would be acquired during such a study would also require months (if not longer) to correlate and understand. Nonetheless, the wealth of information obtained from a very thorough study could perhaps answer many questions regarding the structures of overlapping molecular formulas, but it is more likely that MS/MS will be utilized in the near future for specific target compounds that have been found to be markers for particular samples.

6. Conclusions

FTICR-MS coupled to API sources is clearly a powerful technique for the examination of the complex composition of NOM and has facilitated the overall characterization of NOM from a variety of source materials. The advent of API sources has allowed for the ionization of large, nonvolatile compounds, and its application to the polar, polyelectrolytic NOM mixtures has made a significant contribution to the understanding of the composition and reactivity of NOM. The ultrahigh resolving powers of FTICR-MS, those in the range of $10^5$, can separate and resolve the numerous peaks per nominal mass detected for NOM samples, making it the mass spectrometer of choice for investigations of NOM. Furthermore, the
accuracy of the instrument is capable of determining m/z values to the fifth decimal place, from which molecular formula assignments can be made fairly reliably. The ability to assign formulas to the multitude of peaks detected across the mass range of 200-1000 allows for the characterization of NOM at the molecular level.

In this review, we have described the application of FTICR-MS to the analysis of NOM, discussed some important parameters for sample preparation and the acquisition of high quality data, reviewed some of the recent publications of NOM studies utilizing FTICR-MS, assessed the emergence and importance of recent statistical methods, and evaluated the current limitations of the instrument and subsequent data processing. All things considered, the precision, sensitivity, and ultrahigh resolution offered by FTICR-MS reveals molecular level details of NOM composition, which has transformed our knowledge of NOM chemistry. We are confident that as FTICR-MS technology continues to improve, that this instrumentation will be utilized more widely and will lead to significant advancements in not only the areas of analytical and environmental chemistry, but also for the NOM, chemical oceanography, and mass spectrometry communities.

7. References


Fourier Transform Mass Spectrometry for the Molecular Level Characterization of Natural Organic Matter: Instrument Capabilities, Applications, and Limitations


Kujawinski, E.B.; Freitas, M.A.; Zang, X.; Hatcher, P.G.; Green-Church, K.B. & Jones, R.B. (2002). The application of electrospray ionization mass spectrometry (ESI MS) to the


Fourier Transform Mass Spectrometry for the Molecular Level Characterization of Natural Organic Matter: Instrument Capabilities, Applications, and Limitations


This book aims to provide information about Fourier transform to those needing to use infrared spectroscopy, by explaining the fundamental aspects of the Fourier transform, and techniques for analyzing infrared data obtained for a wide number of materials. It summarizes the theory, instrumentation, methodology, techniques and application of FTIR spectroscopy, and improves the performance and quality of FTIR spectrophotometers.

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