Data visualization approaches for rapid, heuristic interpretation of highly complex mixtures of structurally and compositionally diverse chemical entities associated with pharmaceutically relevant materials

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Visualization vs Imaging

Visualization is a technique and a process of translating information and large datasets into graphic or pictorial representations in order to promote and/or clarify communication of unique features or concepts.

Imaging involves the capture, storage and processing of signals to produce or enhance an image or a visual representation of an object, often in 2D/3D graphical representation (e.g., pixel/voxel rendering).

A *map* is an example of a visualization involving diagrammatic representation or collection of data illustrating relationships, characteristics, distribution, size, number or spatial arrangement of unique features over a given area and according to a chosen scale.

 Maps can be two-dimensional representations of three-dimensional space or can be threedimensional.

Some examples...

Semi-quantitative ...and quantitative maps



Some examples...

A qualitative map: Paris, 1900



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Contain many components across a range of concentrations

- Petroleum products (fuels, polymers, etc.)
- Natural organic matter (dissolved marine organic matter)
- Biological materials (cell lysates, fermentation media, metabolites, etc.)
- Formulated drugs and pharmaceutical materials



... of pharmaceutical relevance typically include:

- >Extractables and leachables
- Related substances in starting materials, intermediates, API and products
- Process reaction impurities, reagents and their reaction products
- Excipients and drug-excipient interaction products
- Degradation products



- Present unique challenges to the analyst
 - Multiple, diverse components at various concentrations
 - Abundant additives and formulation components can complicate the search for analytes present at low levels
 - Higher MW analytes = many elemental composition possibilities
 - Timing for results delivery lengthens with sample complexity
- Require advanced technologies and innovative tools
 - Deep expertise
 - Advanced instrumentation
 - Data reduction and visualization



Additives and Processing Aids in Plastic and Rubber Materials



Molecular features contributing to compound diversity

Structural diversity

- Degree of unsaturation (rings/double bond equivalents)
- Substituents
- Functional groups

Isomeric diversity

- Structural isomers
- Positional isomers
- Geometric isomers
- Stereoisomers

Compositional diversity

- Heteroatom content and type
- Isotopic contributions



Polymer Supply Chain for Pharmaceutically Relevant Materials



How complex is "complex" for pharmaceutically relevant materials?

Chromatographic complexity



Mass spectral complexity



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Mass spectral complexity



Mass spectral complexity



To Extol the Virtues of Accurate Mass Measurement...



Correlating Structure with Formula × Ø Draw Q,

SciFinder'

Substances -

Edit Search Enter a query..



Some Tools for Visualizing Complex Molecular Mixtures: Kendrick Mass Defect Diagram

Kendrick mass defect diagram (2D or 3D modified)

- Rescales IUPAC mass scale (based on 12C mass = 12.000 00 Da) to "Kendrick" mass scale for CH2 (e.g., converts CH2 mass from 14.015 65 to 14.000 00 Da)
- Sorts compounds into homologous series by compound "class" (numbers of O, N, other heteroatom)
- Compound type (rings and double bonds) and degree of alkylation (CH₂ groups)



Some Tools for Visualizing Complex Molecular Mixtures: van Krevelen Diagram

van Krevelen diagram (2D or 3D modified)

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- Projects elemental composition according to H/C and heteroatom/C atomic ratios
- Compounds are classified according to degree of saturation, alkylation, heteroatoms, dehydration, deamination
- Compositional differences are amplified permitting classification, prediction of composition and origin



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van Krevelen Diagram: Visualizing Relationships

FT-ICR LC-MS analysis of IPA extracts (5 materials) – mapping extractables



 A traditional <u>discrete</u> spectral library is a collection of *individual* spectra representing individual chemical entities

Each library entry represents a single, *unique compound*



A <u>spectral map</u> library is a collection of sets of spectra representing visualizations of complex mixtures of structurally and compositionally diverse chemical entities

Each library entry represents a *unique extractables profile* corresponding to a component or material



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Standardized visualizations saved as QR Codes represent unique, highly complex mixture and material "fingerprints" that can be compared both qualitatively and quantitatively



Pairwise or batch comparison with a reference material and index of similarity / dissimilarity with anomaly detection



High resolution digitized image grouping to evaluate change over time (pixelated 2-D or voxelated 3-D image)



LC-MS Analytical Workflow for Extractables Assessment



"Mass superspectrum"

Difference Correlation and Self-Calibrating Mass Maps

- Detect monoisotopic peaks
- Determine charge state
- Generate an *autocorrelation spectrum* using Fourier transform methods
- Autocorrelation spectrum has peaks corresponding to *m/z differences* source spectrum
- Apply Kendrick mass defect methods to classify peaks based on formula relationships
- Group related compositions using hierarchical clustering or automated van Krevelen plot analysis





Difference Correlation and Self-Calibrating Mass Maps



Identify monoisotopic peaks, charge state and

accurate m/z difference correlation

	Formula	.055	Calc Mass	Expt Mass	Mono Inty	Abs Error	mDa Error	ppm Error	
1	H2O		18.0106	18.0106	675963072	0.000055	0.06	3.06	
2	со		27.9949	27.9949	988126208	0.000013	-0.01	-0.47	
3	C2H4		28.0313	28.0313	8958516224	0.000010	-0.01	-0.37	P
4	CH2O2		46.0055	46.0054	5211857920	0.000071	-0.07	-1.55	
5	C3H4O		56.0262	56.0261	1176259584	0.000088	-0.09	-1.57	
6	С3Н6О2		74.0368	74.0367	7576710656	0.000065	-0.07	-0.88	
7	С5Н8О		84.0575	84.0574	799050624	0.000101	-0.10	-1.20	
8	C4H6O3		102.0317	102.0316	552958464	0.000080	-0.08	-0.79	
9	C5H10O2		102.0681	102.0680	5281296384	0.000079	-0.08	-0.78	
10	C4H8O4		120.0423	120.0421	1193746816	0.000126	-0.13	-1.05	
11	C6H10O3		130.0630	130.0629	842935104	0.000121	-0.12	-0.93	
12	C7H14O2		130.0994	130.0993	1729864064	0.000091	-0.09	-0.70	
13	C8H14O3		158.0943	158.0941	606668480	0.000169	-0.17	-1.07	

Basis for a *self-calibrating* mass map

Difference Correlation and Self-Calibrating Mass Maps

N-dimensional difference correlation and self-calibrating mass maps



FT-ICR LC-MS total ion chromatograms (right) and mass spectra (left) of IPA extracts of multi-layer biopharmaceutical bags from different sources





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Comparative Kendrick mass defect diagrams



Reference Map

Comparative 3D van Krevelen diagrams





Nonionic Surfactants: Comparative Control and Stress Degradation Samples



Nonionic Surfactants: Similarity Index and Anomaly Detection



Elastomers: Similarity Index and Anomaly Detection



Elastomers: Similarity Index and Anomaly Detection



In Conclusion, Visualization Approaches Enable...

- Simple, intuitive approach to enabling rapid "at-a-glance" heuristic assessment and visualization of highly complex mixtures of structurally and compositionally diverse chemical entities
- Graphically sorting complex mixture components by series or class based on structural and compositional features
- Confident assignment of elemental composition to higher masses with initial assignment of elemental composition to a limited number of lower MW components
- N-dimensional self-calibrating mass maps based on mass difference correlation
- Rapid and comprehensive pairwise and batch comparison with an index of similarity and anomaly detection
 - Change management
 - Failure mode analysis
 - Material characterization and selection

Acknowledgements

Wolfgang Mueller Tim Baker (Procter & Gamble) Bob Strife (Sierra Analytics) David Stranz (Sierra Analytics) Scott Campbell (Sierra Analytics) George Maydwell (Sierra Analytics) Rob Smith (Prime Labs) Diane Paskiet (West Pharmaceutical Services) Cheryl Stults (C & M Technical Consulting) Ryan Rodgers (NHMFL/FSU) Alan Marshall (FSU) Daniel Norwood (Feinberg Norwood & Associates Pharma Consulting)

Thank you!

Many thanks to the PQRI Secretariat and the PQRI/FDA 2021 Conference Organizers!