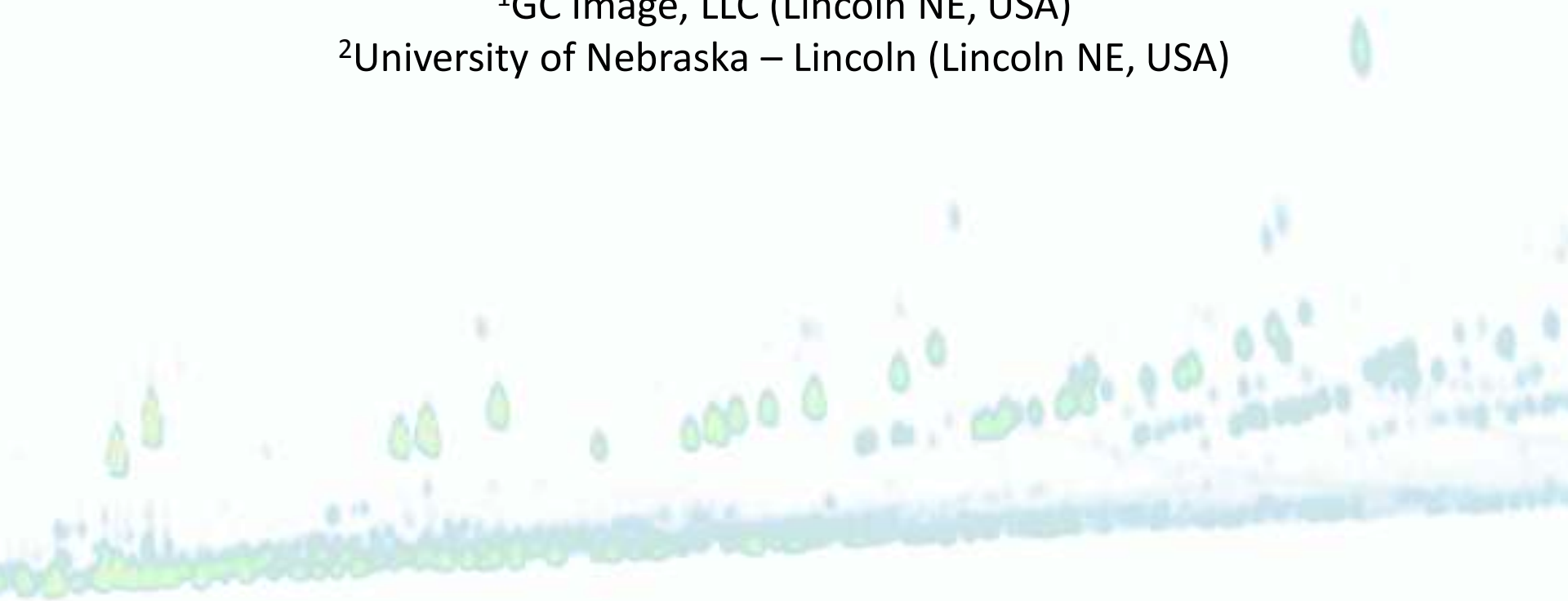


Advanced Software Tools for Plant Substances Analyses using Comprehensive Two-Dimensional Chromatography with High-Resolution Mass Spectrometry

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Outline

- Introduction
- Multi-Sample Analysis: Workflows and Challenges
- Automated Workflow for Non-targeted Multi-sample Analysis
- Software Tools for Identifying Biomarkers with High-Resolution Mass Spectrometry
- Conclusions

Introduction: Goals

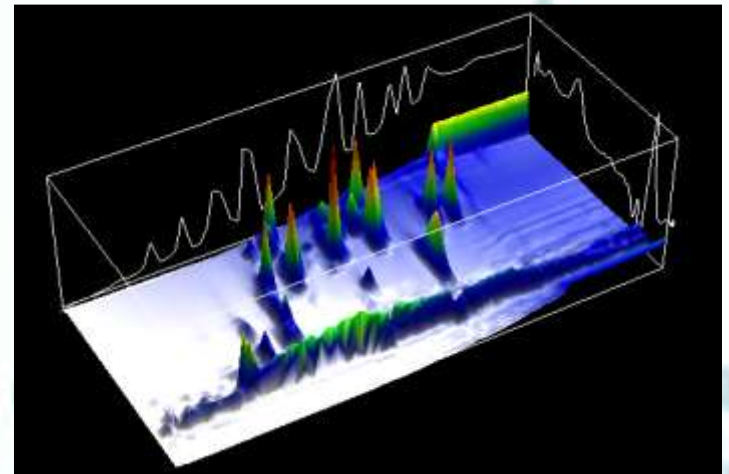
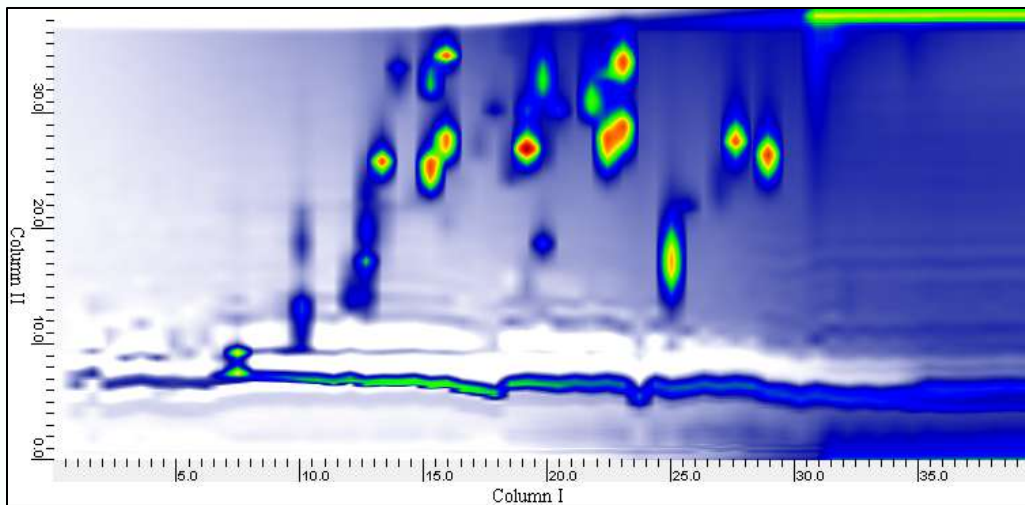
- Goal: Understand the states or processes of biological systems of plants
 - Discover & document compounds/metabolites with different relative compositions among sample classes or various stages as potential biomarkers
- Effective Analytical Methods:
 - Open, unbiased, and comprehensive
 - Able to analyze highly complex mixtures of compounds

Introduction: Goals

- Comprehensive two-dimensional gas and liquid chromatography (GCxGC and LCxLC)
 - Much greater separation capacity and signal-to-noise than traditional one-dimensional chromatography
 - High sensitivity and selectivity when coupled with high-resolution mass spectrometry (HRMS)
 - Produces large, highly complex data that is challenging to analyze.

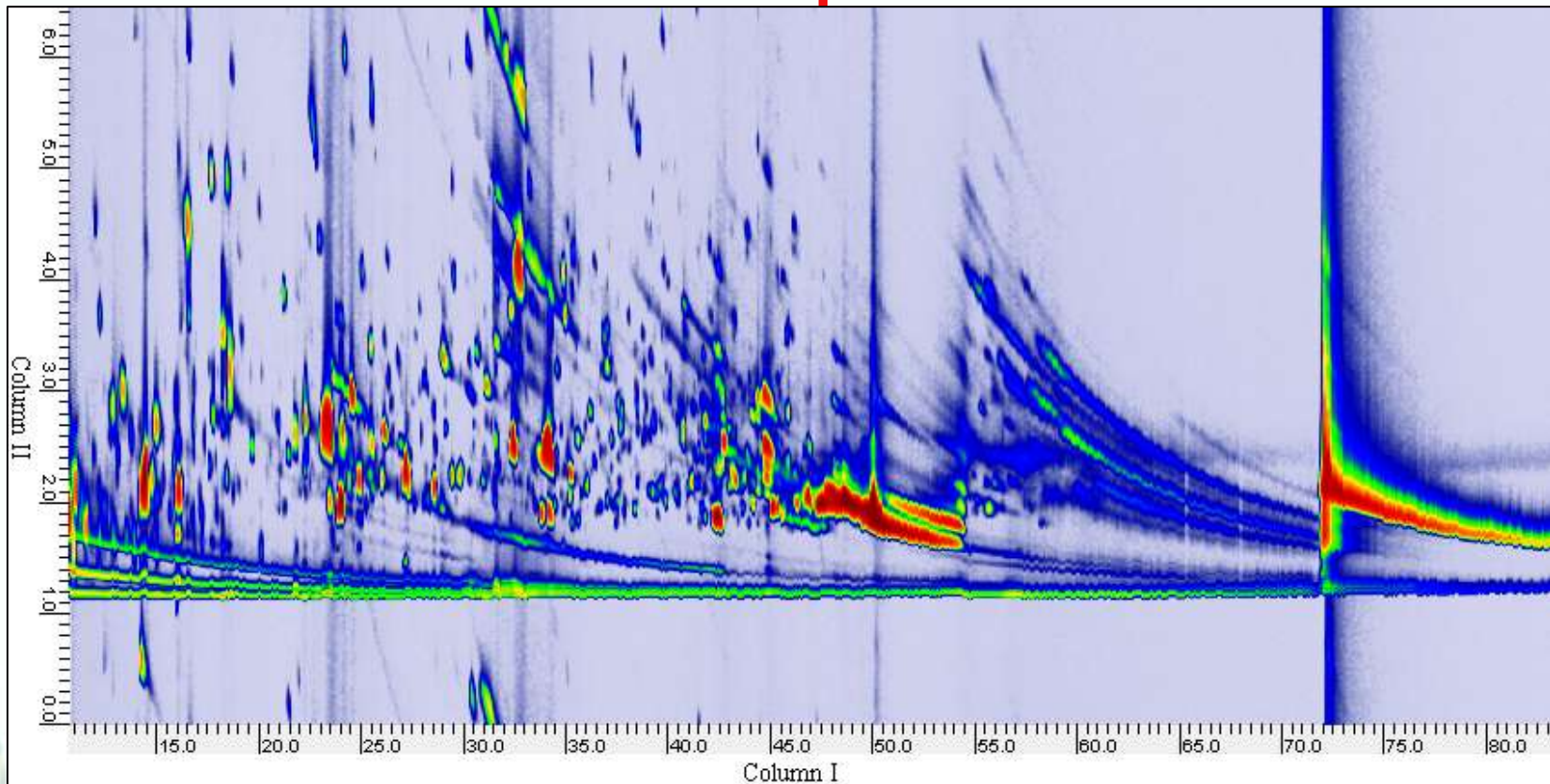
Introduction: Two-Dimensional Chromatogram

- Peaks are two-dimensional, with several Column II chromatograms cross each peak.



A LCxLC chromatogram of a standard mixture of polyphenolic compounds acquired by Agilent 1290 Infinity 2D-LC (S.E. Reichenbach and E. Naegele. Agilent Application Notes, 2013).

2D Chromatogram: Another Example



A GCxGC chromatogram of a wild type strain of rice blast fungus acquired by Agilent 7890B/ZOEX ZX2 thermal modulation system coupled with Agilent 7200 Q-TOF (Sofia Aronova *et al.*, ISCC 2014)

Introduction: Data Analysis

- Challenge: Comprehensive analysis of many compounds from multiple chromatograms
- Requirements:
 - Effective chromatography
 - ***Effective data processing***
 - ***Effective multi-sample alignment and analysis***
 - ***Automation***
- Use informatics from 10+ years of R&D by GC Image.

Outline

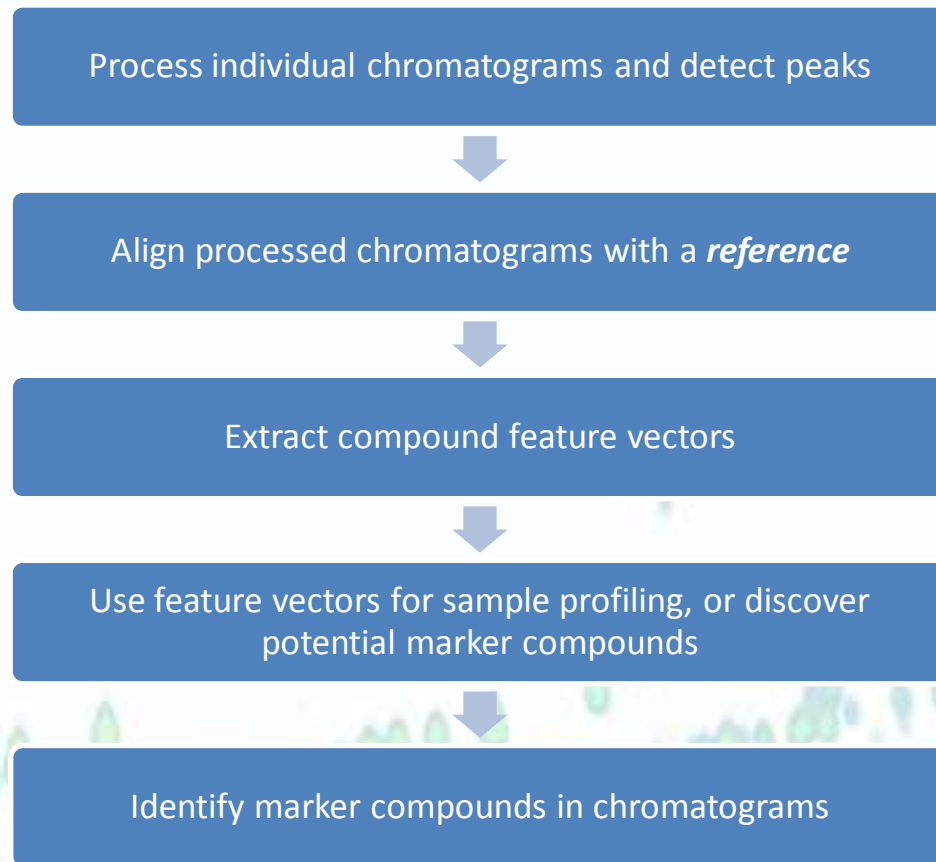
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Multi-sample Analysis

- **Applications:**

- **Clustering** – Discover sample subsets, such that samples in the same subset are similar and samples in different subsets are dissimilar.
- **Change Detection** – Discover uncharacteristic differences, progressive trends, or cyclical patterns in a sample sequence.
- **Classification** – Given a training set of labeled samples from multiple classes, discover the class of an unlabeled sample.
- **Chemical Fingerprinting** – Given a set of samples from known sources, discover the unknown source of a test sample.
- **Biomarker Discovery** – Given a set of labeled samples from multiple classes, discover the features that are most salient for distinguishing the classes.

Multi-Sample Analysis: Workflow



Multi-Sample Analysis: Targeted Analysis

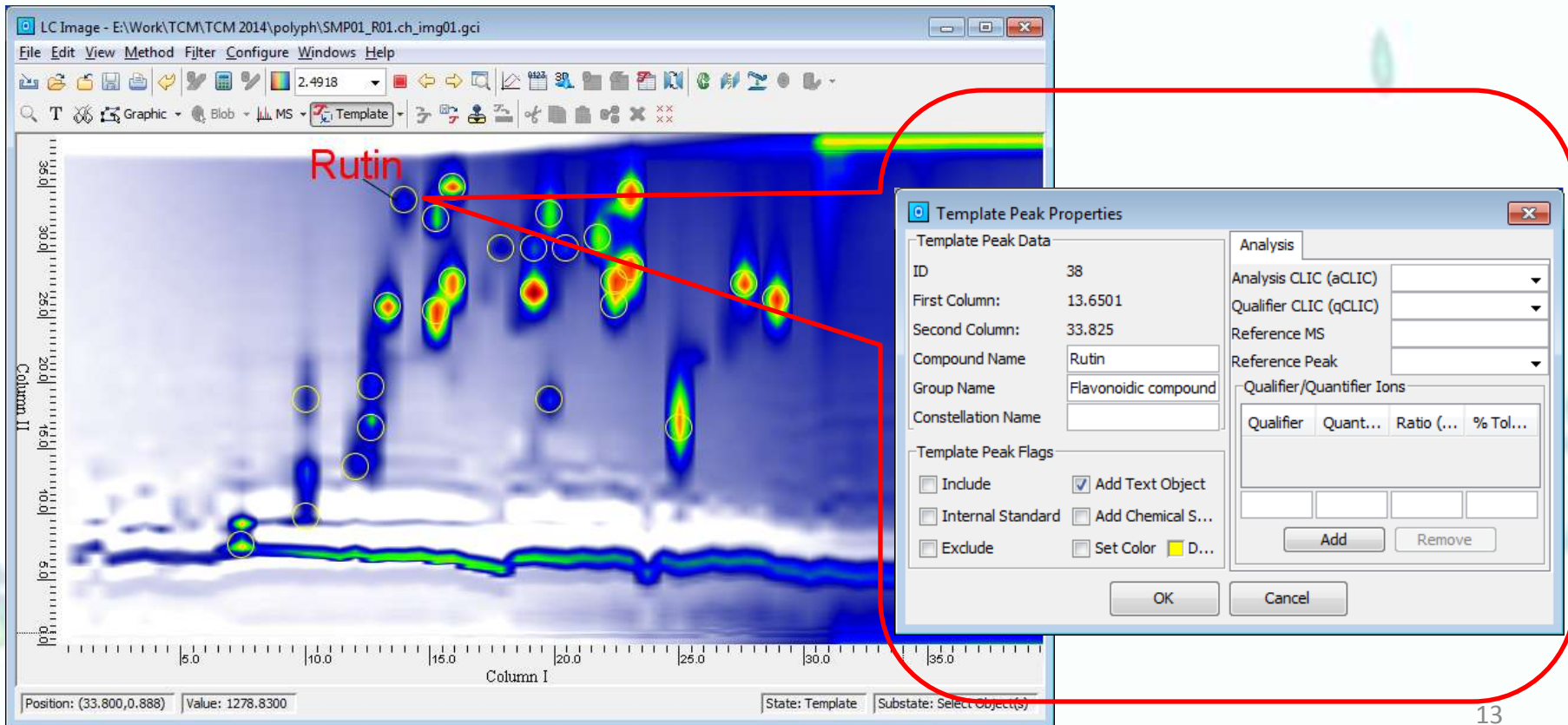
- Reference
 - A standard mixture of known compounds
 - A selected reference sample
- Advantages:
 - Peak identification can be optimized
 - More efficient
- Disadvantages: Limited to targeted compounds

Targeted Analysis: Template Matching

- Template Matching from GC Image:
 - A powerful tool for automated identification
 - Use advanced pattern recognition to identify peak pattern in a new chromatogram
- A template:
 - Peak patterns (including RTs and spectra)
 - Chemical logic expressions for peak matching constraints & quality assurance (QA) assessment
 - Other metadata, e.g., groups & descriptive annotations

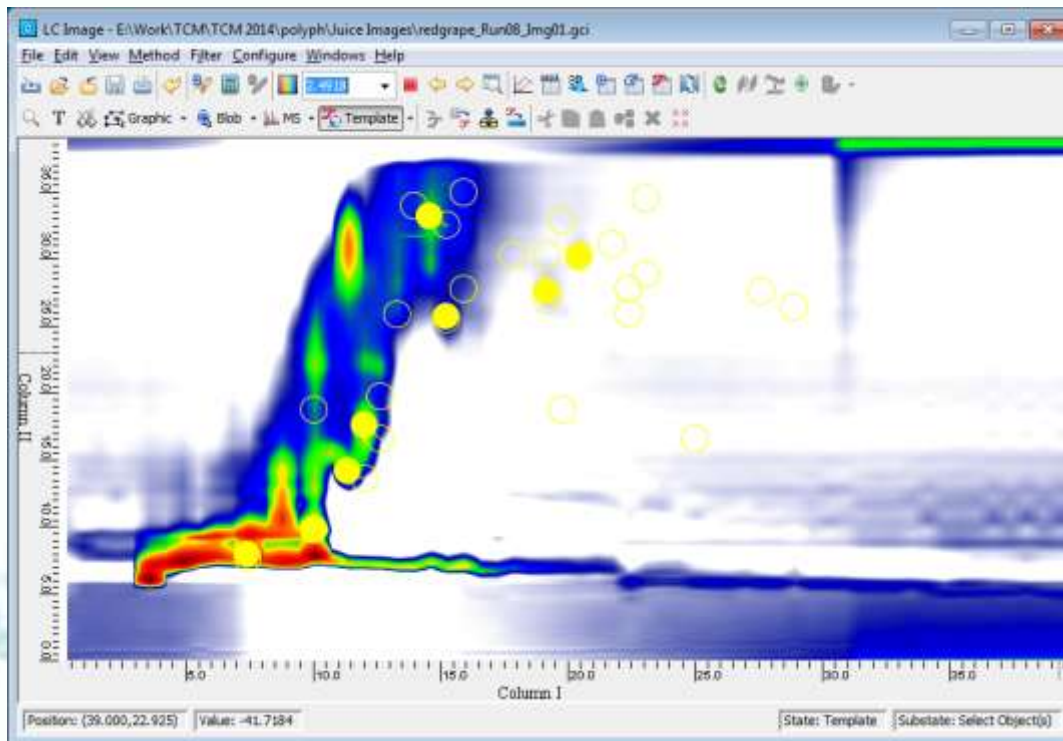
Template Matching: Example

- A template that contains 26 polyphenolic compounds from the standard mixture



Template Matching: Example

- Match the template to a grape juice sample.
 - Some peaks are matched shown as solid circle
 - Others are unmatched:



- Not exist
- Undetected
(Trace or co-eluted peaks)
- Unmatched
(RT or spectral mismatch)
- May mismatch
(nearer peak)

Multi-Sample Analysis: Non-targeted Analysis

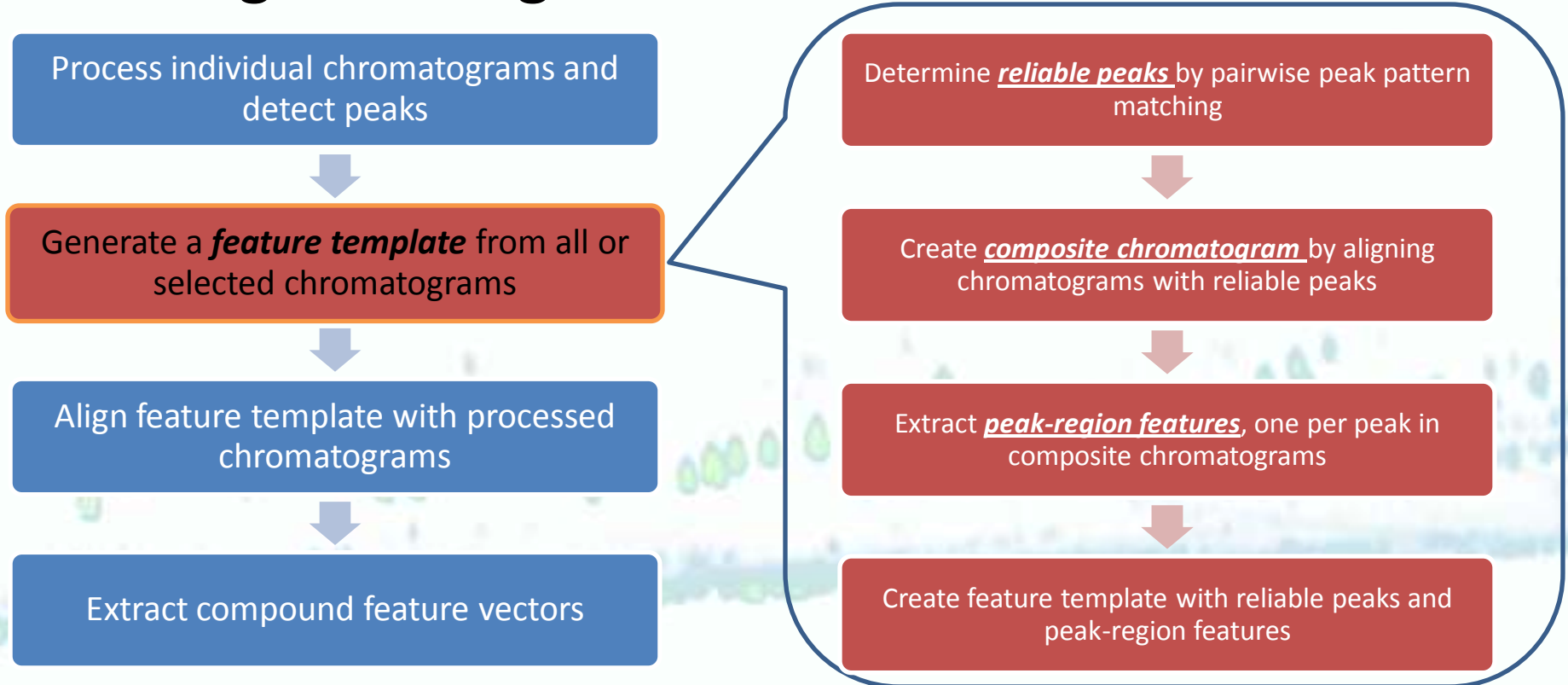
- **Non-targeted analysis** requires features to characterize all compounds, not just targeted or selected compounds.
 - Characterize with retention times, intensity, mass spectrum.
- **Multi-sample analysis** requires matching features across all samples, providing “apples-to-apples” comparisons.
 - Can be difficult for complex, information-rich two-dimensional chromatography data.
- **Non-targeted multi-sample analysis** requires matching all features across all samples, and so are the most challenging.
- What will be the reference?

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Non-targeted Multi-sample Analysis

- Automated workflow supported by GC Image's Image Investigator™ software:



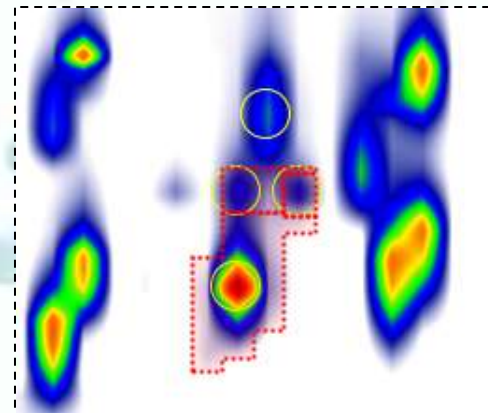
Feature Template: Reliable Peaks

- Reliable peaks are determined from the bidirectional pairwise matching of all possible pairs of chromatograms (Reichenbach *et al.*, *Anal Chem*, 85:4974, 2013).



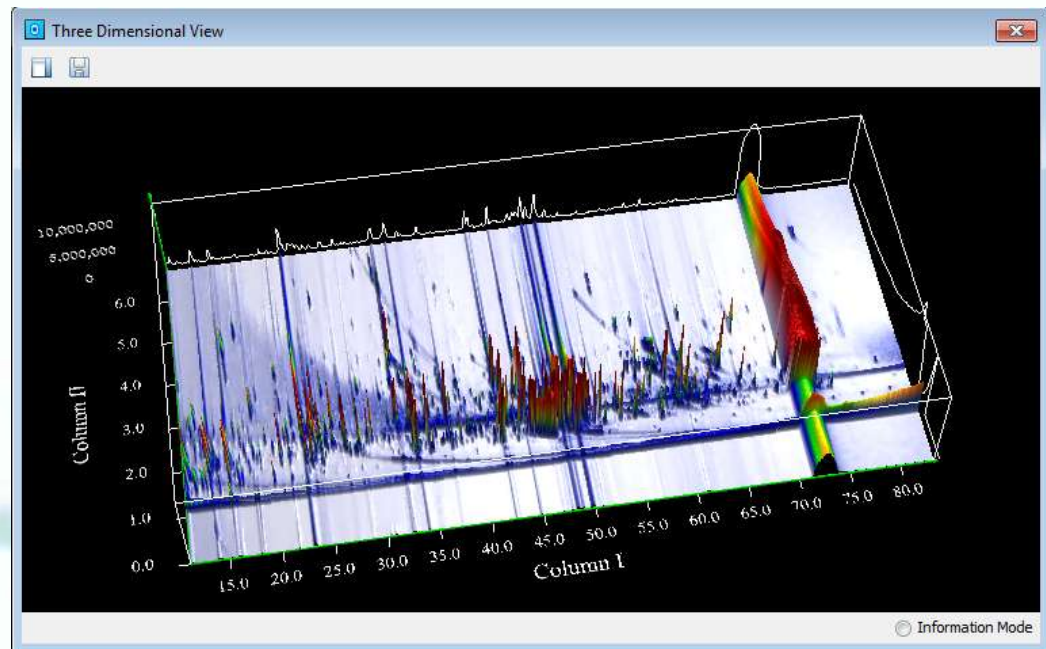
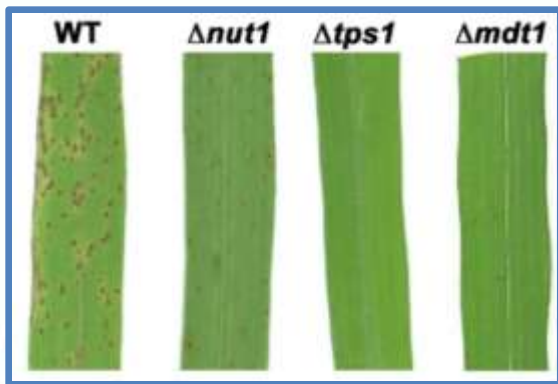
Feature Template: Peak-Region Features

- Goal: Define a region for every peak in every chromatogram
 - **Composite Chromatogram:** All chromatograms are aligned and combined (e.g., by addition) to form a single composite chromatogram that is reflective of all of the constituents in all samples.
 - Peak-region features are delineated by peak detection in the composite chromatogram (Reichenbach et al., J Chromatogr A, 2012)
- Peak-Region features are comprehensive, accounting for every analyte, and feature matching is implicitly performed by the retention-time mapping.



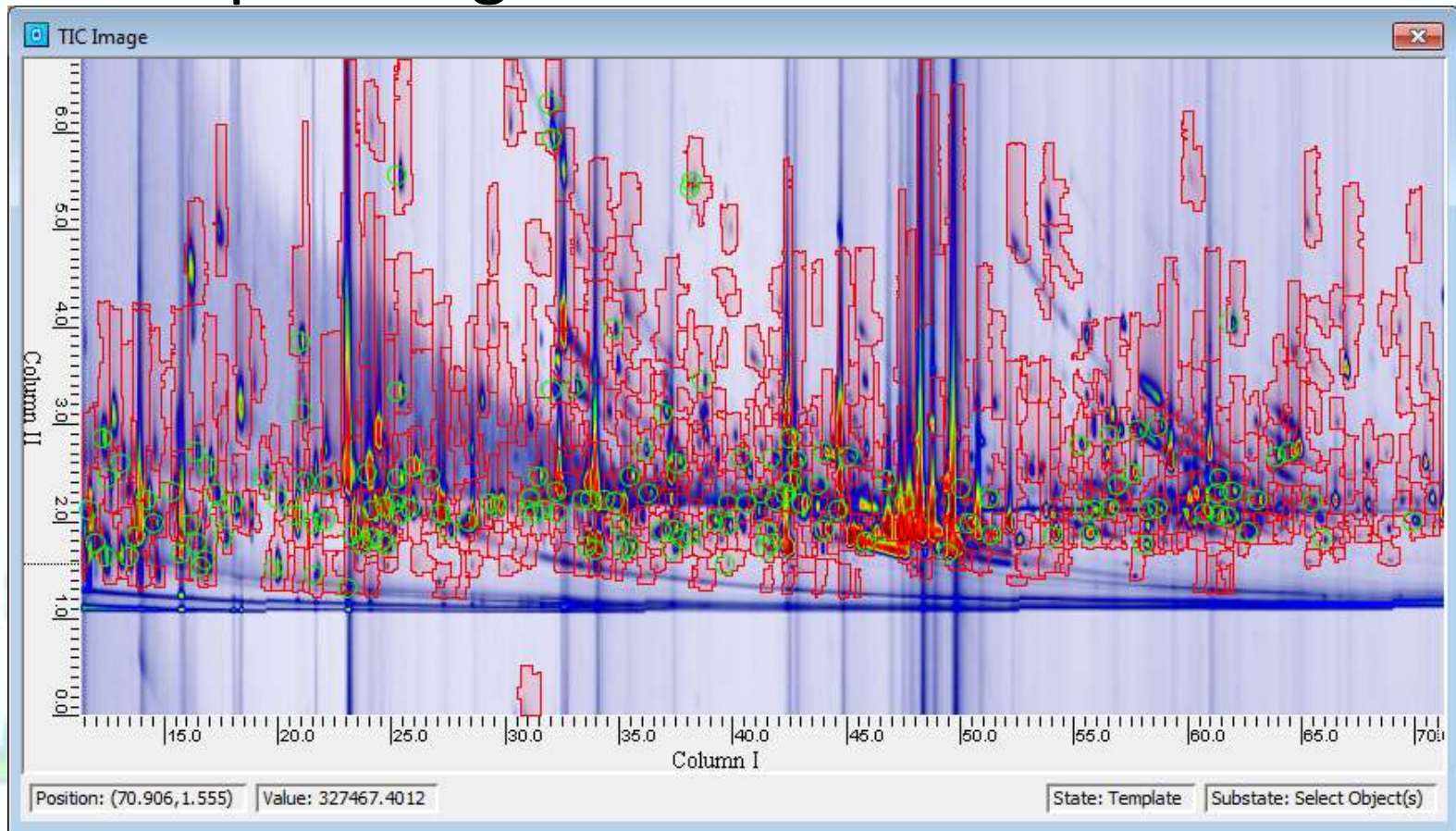
Example Data

- Non-targeted analysis of 4 types of rice blast fungus *Magnaporthe oryzae* (Sofia Aronova *et al.*, ISCC 2014)
 - Instrument: Agilent 7890B/ZOEX ZX2 thermal modulation system coupled with Agilent 7200 Q-TOF
 - Data analysis software: GC Image Pre-Release 2.5a0



Data Analysis: Feature Template

- The feature template with 159 reliable peaks & 572 peak-region features

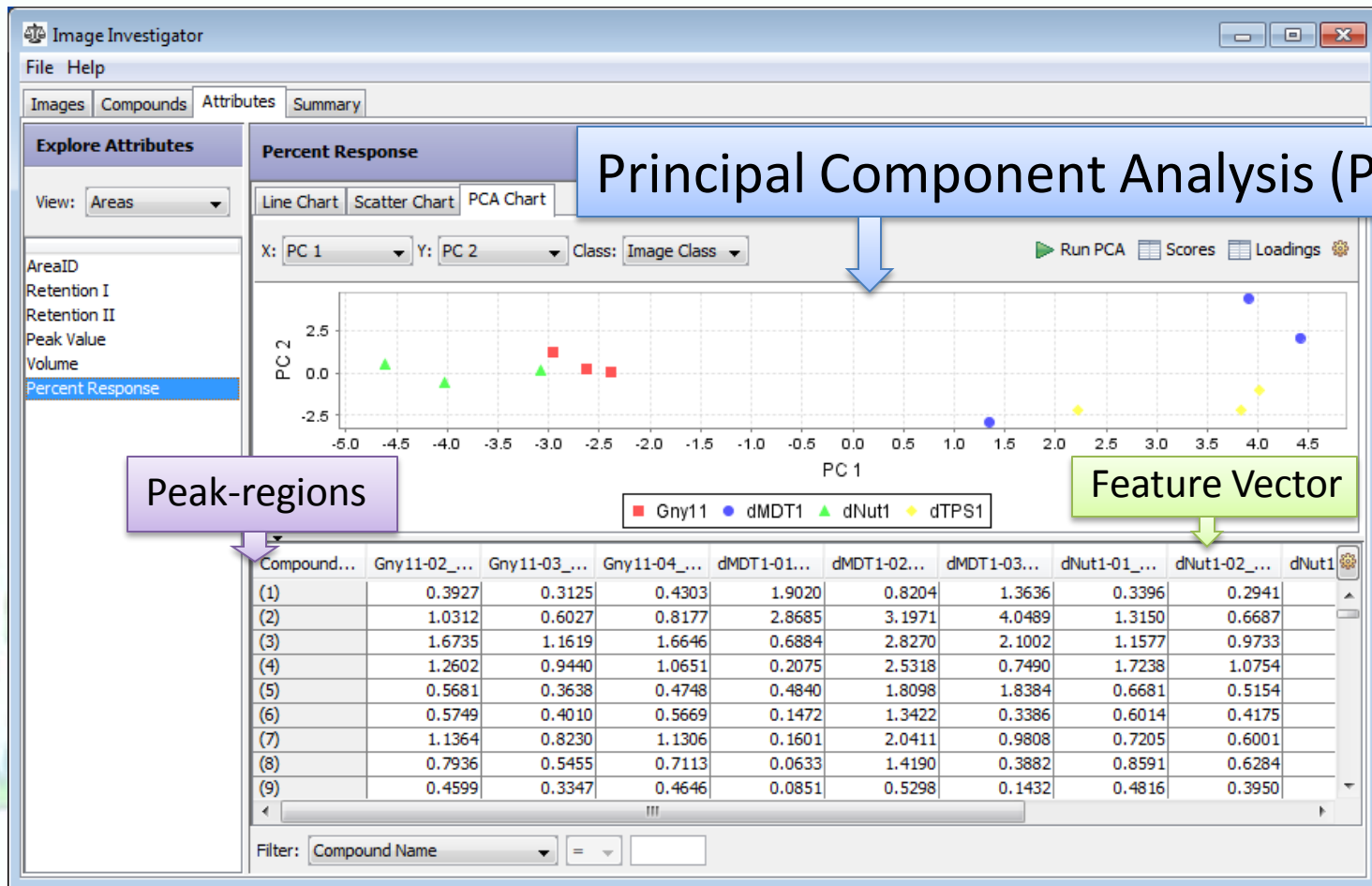


Data Analysis: Feature Measures

- Analyze all features in each chromatogram:
 1. Match feature template to each chromatogram
 - a. Match pattern of reliable peaks to detected blobs
 - b. Geometrically adjust peak-region features relative to matched peaks
 2. Record feature vector with quantitative attributes for each peak-region feature in each chromatogram
- Typical attributes for each feature vector
 - Retention times, retention index
 - Volume (TIC in peak-region)
 - Percent response (volume / Σ volumes)

Data Analysis: Feature Vectors

- Each chromatogram has a feature vector with attribute values of all peak-regions



Data Analysis: Feature Statistics

- Use attribute statistics (e.g., on previous slide) to select features of interest:
 - Per peak-region over all chromatograms: Mean, standard deviation, & RSD
 - Per class, per peak-region: mean, standard deviation, & RSD
 - Class-to-Class & Class-to-Others mean differences
 - Multiclass Fisher Linear Discriminant Ratio
 - Class-to-Class & Class-to-Others Fisher Ratios

Data Analysis: Feature Selection

- Potential biomarkers:

- Large F value among all classes

$$F(x_1, \dots, x_K) = \frac{\sum_i N_i (\mu_i - \mu)^2 / (K-1)}{\sum_{i,j} N_{i,j} (x_{i,j} - \mu_i)^2 / (N-K)}$$

- Larger Fisher ratio between one class and others

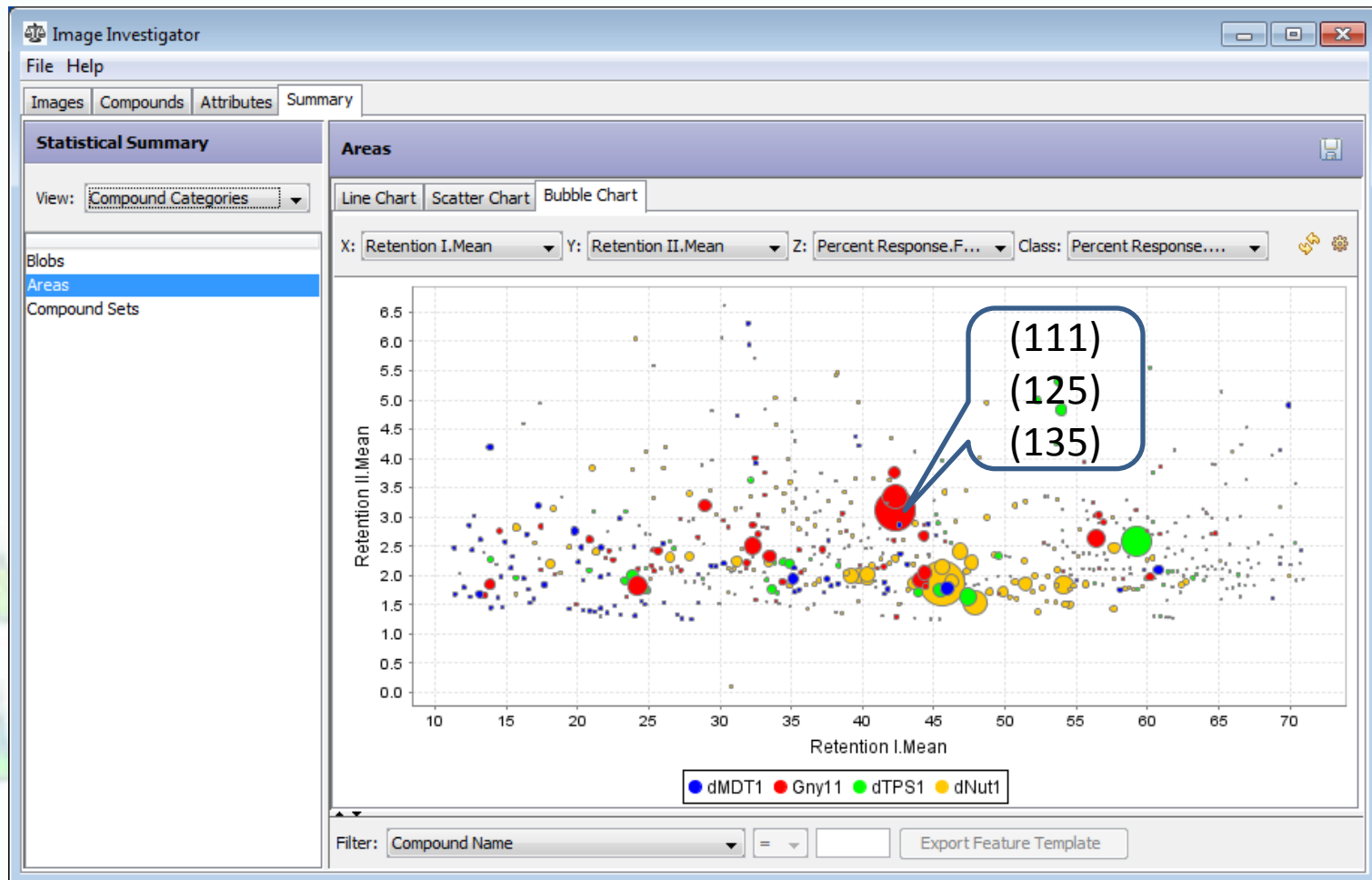
$$FDR(x_1, x_2) = \frac{(\mu_1 - \mu_2)^2}{(\sigma_1^2 + \sigma_2^2)}$$

- Both measures assess between-group variance against within-group variance.

- Analyze peak-regions that have large F values and Fisher ratios as prospective biomarker using HRMS

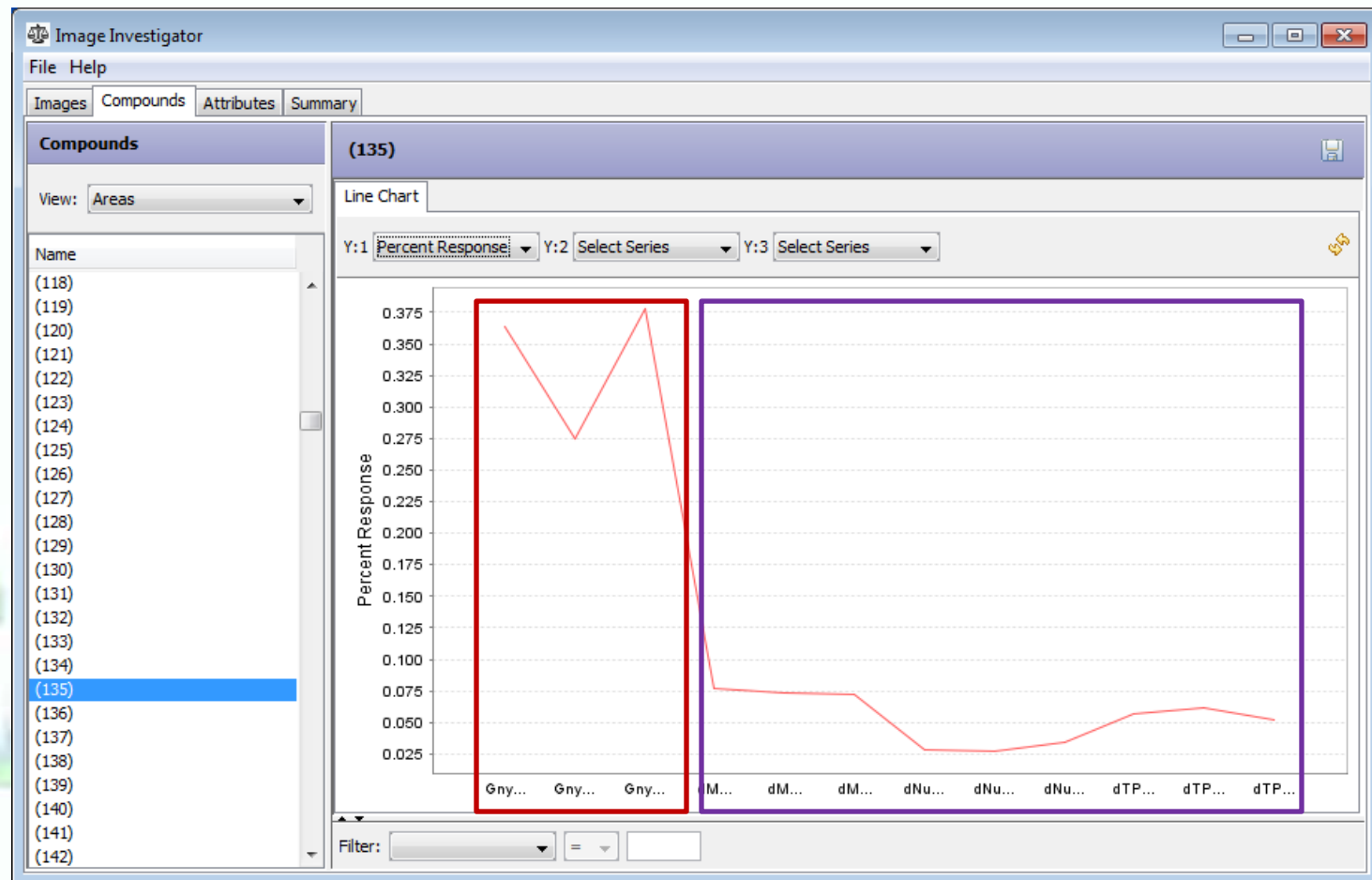
Data Analysis: Selected Features

- Several features with larger F value for all samples and larger Fisher ratios between one type against other types of samples



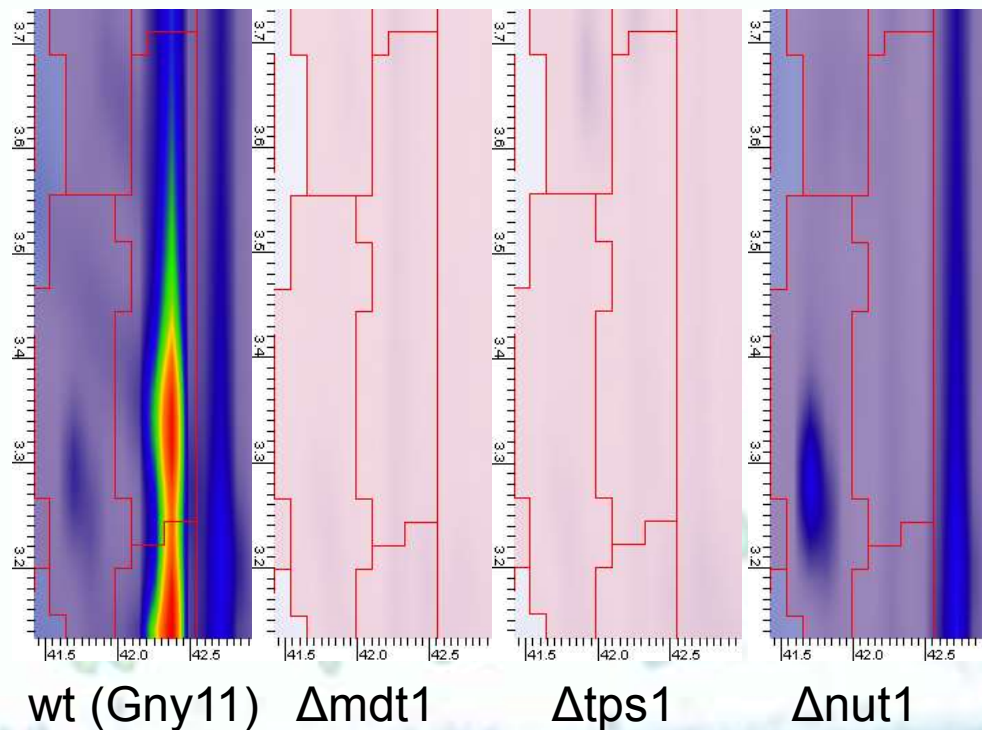
Data Analysis: Selected Features

- Between-class differences in peak-region %response also indicates prospective biomarker



Data Analysis: Selected Features

- Chromatograms of Peak-Region (135) for the four classes:

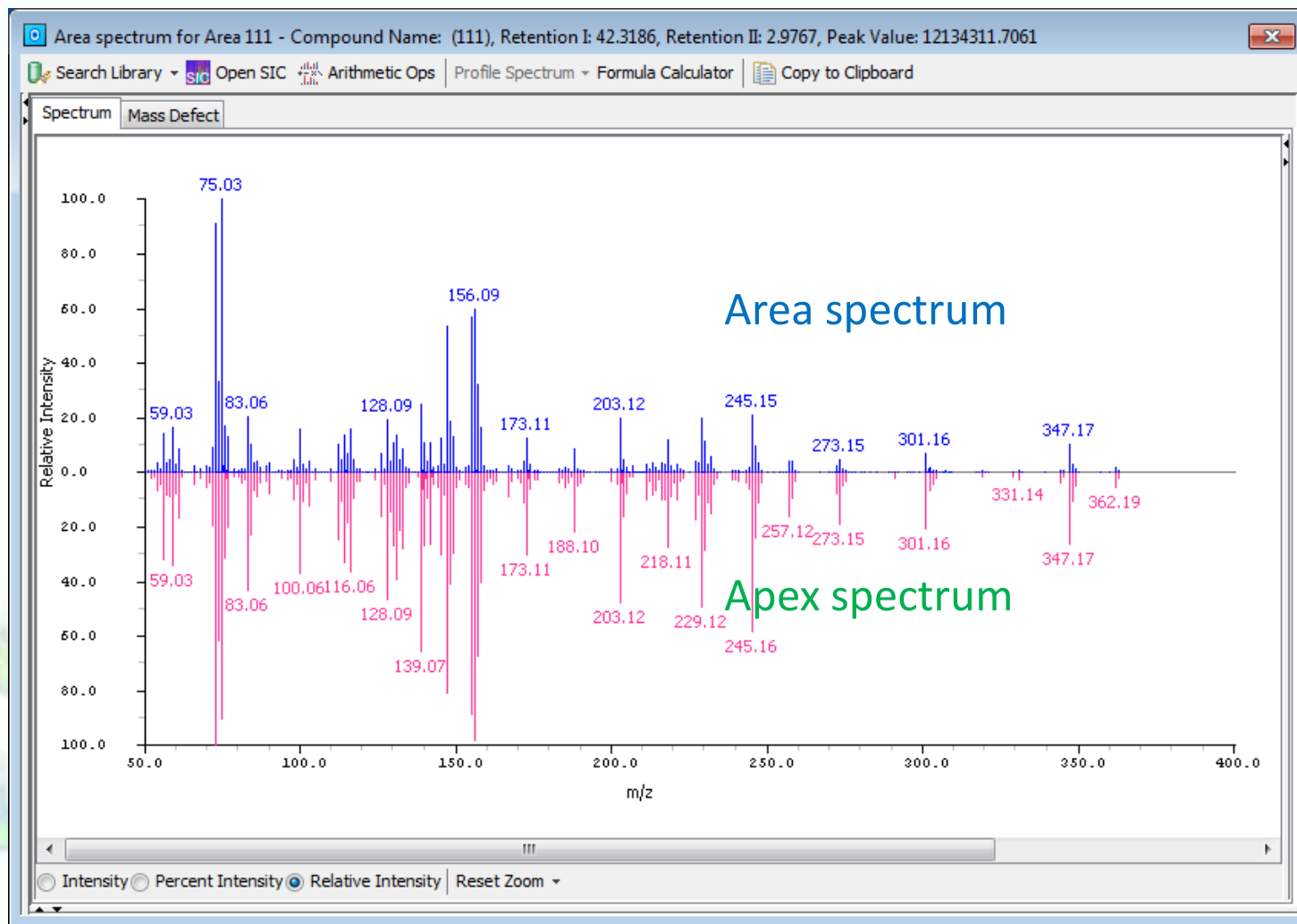


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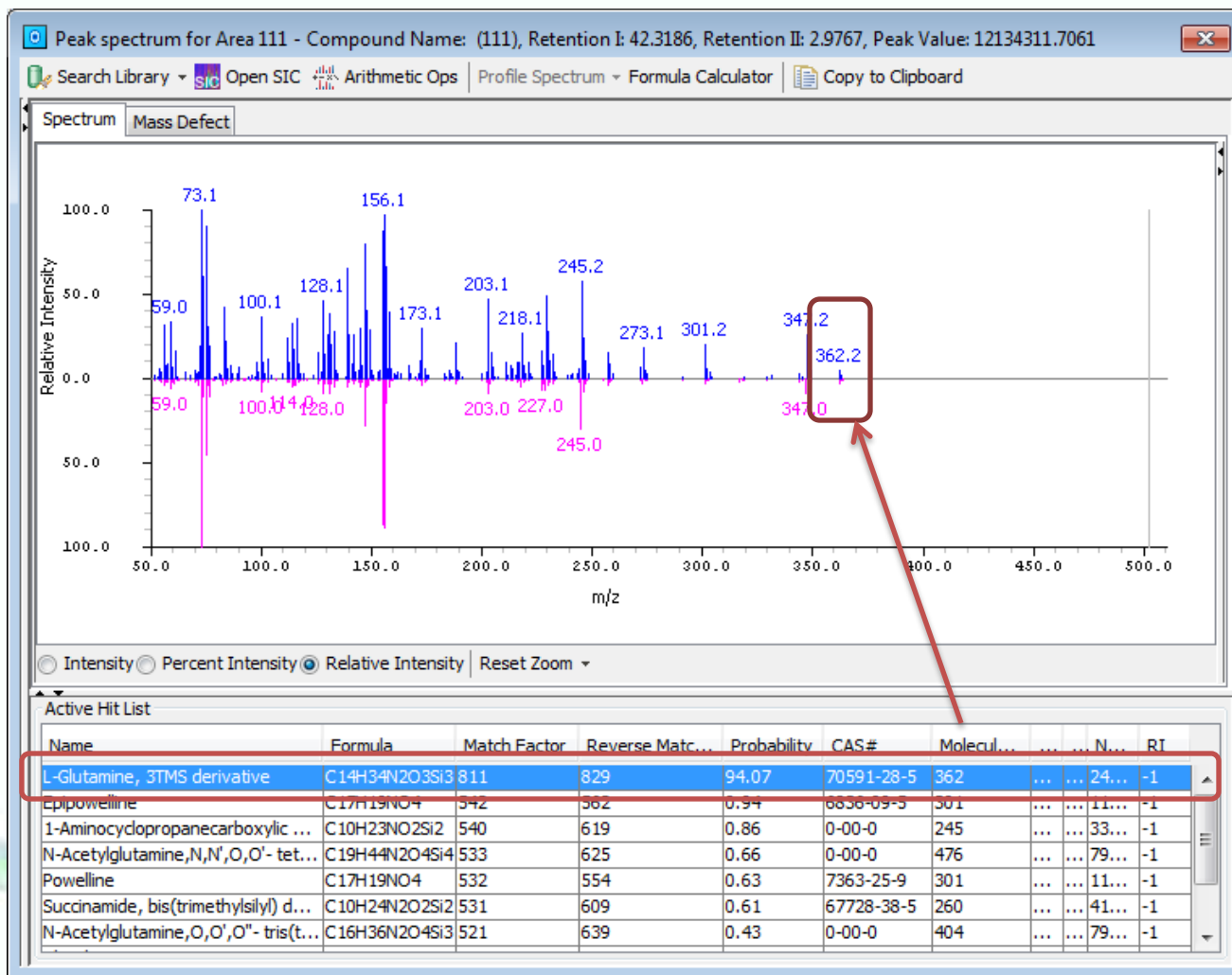
HRMS Analysis: Peak-Region (111)

- Good spectral purity, e.g., Area & apex spectra



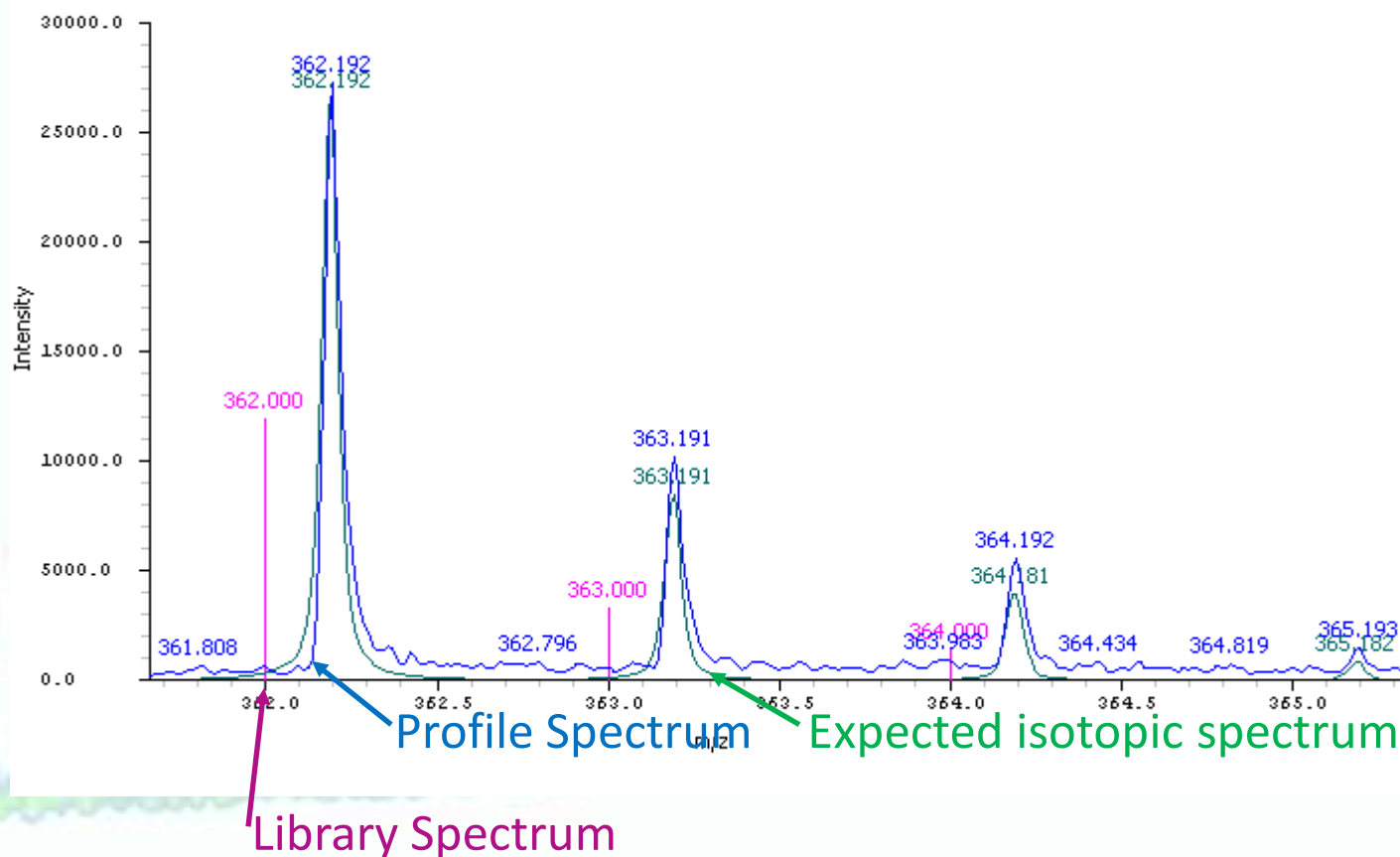
HRMS Analysis: Peak-Region (111)

- NIST14 MS match (811) & reverse match (829)



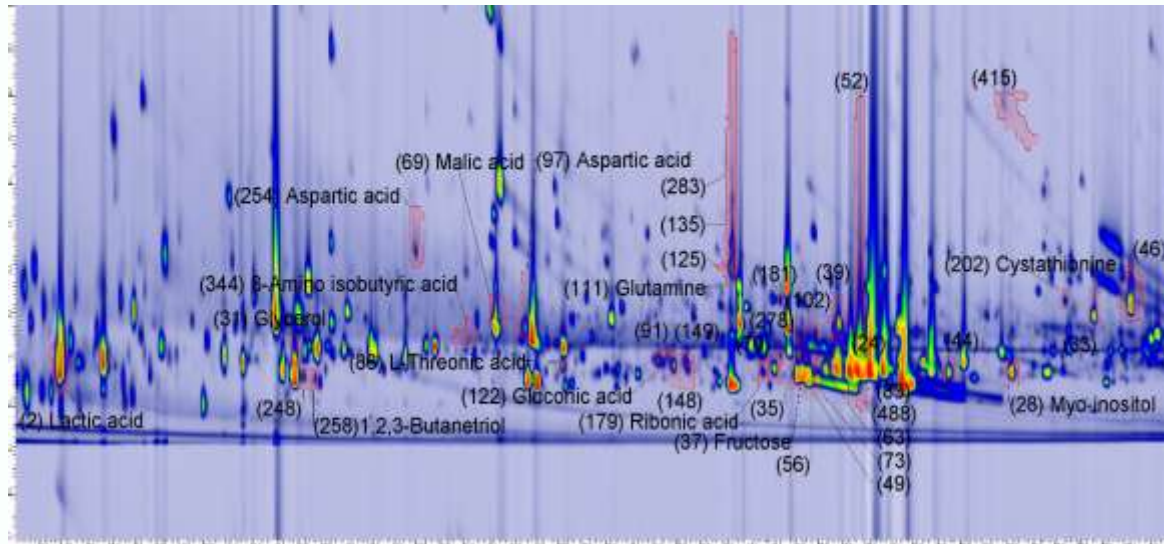
HRMS Analysis: Peak-Region (111)

- Confirm the identification with exact mass
 - L-Glutamine, 3TMS derivative, $C_{14}H_{34}N_2O_3Si_3$



HRMS Analysis: Peak-Region

- Prospective peak-region biomarkers (with red polygons), labeled with ID and compound name (if identified).



- Not all compound identities can be determined
 - Not in NIST library (or other libraries, including Wiley 8E, Fiehn, Golm)
 - Large molecule with more complex elemental composition
 - Smaller concentration, so smaller ion peaks

Conclusions

- Advanced software tools with comprehensive peak-region feature analysis can effectively detect compounds that are highly differentiated between classes as potential biomarkers.
- Two-dimensional chromatography with HRMS provides superior basis for compound identification, but unknown compound identification remains highly challenging.

Acknowledgements

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